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AUBURN UNIVERSITY

**TERMINAL GUIDANCE AND
NAVIGATION FOR COMET
AND ASTEROID RENDEZVOUS**

FINAL REPORT

Prepared for

**NATIONAL AERONAUTICS AND SPACE ADMINISTRATION
GEORGE C. MARSHALL SPACE FLIGHT CENTER**

Under Contract NAS8-30248

June 15, 1975



**ENGINEERING EXPERIMENT STATION
AUBURN UNIVERSITY
AUBURN, ALABAMA 36830**

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TERMINAL GUIDANCE AND NAVIGATION FOR COMET
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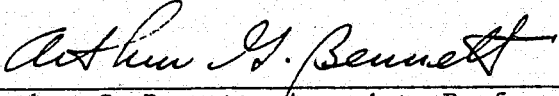
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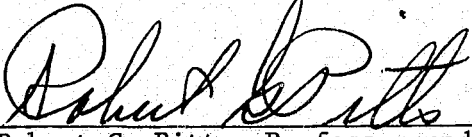
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ABSTRACT

A terminal guidance and navigation scheme developed in earlier work is modified and evaluated for a solar electric propulsion rendezvous mission to comet Encke. The scheme is intended for autonomous, on board use. The guidance algorithm is based in optimal control theory and minimizes the time integrated square of thrust acceleration. The navigation algorithm employs a modified Kalman filter set in measurement variables. Random sequences were generated to simulate measurement errors and the evaluation was conducted with detailed numerical computations which include actual motions of spacecraft and comet. The evaluations showed that the scheme attains rendezvous and maintains station after rendezvous within less than 10 km for estimated "best" measurements and within less than 100 km for estimated "worst" measurements. The measurements required are angles, range and range rate. Angles and range appear to be absolutely necessary. Range rate is not as strong a measurement type and further modifications of the filter will quite probably allow a scheme that does not require the rate measurements.

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1.0 INTRODUCTION

In previous work on the comet and asteroid navigation and guidance problem, an optimal control theory guidance algorithm was devised.¹ An evaluation of this algorithm combined with a Kalman filter method for navigation showed that onboard guidance and navigation is possible with reasonable limits on measurement errors.² But, this evaluation did not fully establish practicality of the onboard approach. Questions of how many measurement types are required and the effects of range of measurement accuracy were not answered. And, the particular coordinate system employed (range and direction cosines) gave unacceptable singularities near coordinate planes. Also, there was the not unusual question of management of state estimate divergence caused by the use of a linear filter with a nonlinear system.

The objective of the work presented here was to carry out further algorithm development to provide first answers to the questions above. Results were to include the precision attainable with different instrument types and accuracies and the onboard computer requirements necessary to implement the scheme.

After initiation of the work, it became evident that the conversion of the previously developed GANDER guidance and navigation computer program to a non-singular coordinate system would be a larger task than originally planned. The divergence question also presented unforeseen difficulties. It was therefore agreed to reduce work on less important tasks. Specifically, thrust errors were simulated by statistical terms

in the filter equations and not by a second method wherein thrust errors would be statistically developed and handled as new state variables. Similarly, the use of available angular measurements to obtain improved estimates of target ephemeris was not investigated. These refinements, properly done, can only improve results. Since successful rendezvous was obtained without these refinements, the decision to drop them was correct.

Identification of specific instruments, their accuracies and power requirements could not be carried out because the requisite data has yet to be developed in the frequency ranges of interest for onboard radar in free space. A request to do this data development could not be funded, so these questions remain open.

2.0 REFORMULATION OF THE GUIDANCE AND NAVIGATION SCHEME

A first step in the work reported here was to eliminate the difficulties encountered with the range and direction cosine coordinates used previously in the Kalman filter. The difficulty arose because of the basic relation among direction cosines that the sum of their squares is unity. Near a coordinate plane one of the direction cosines approaches zero and small numerical differences such as arise from measurement errors can lead, in numerical computation, to imaginary values for this cosine. If this happens, the computer becomes most upset and reports in displeasure with a long string of error statements. There are ways to get around this problem, but in doing so some of the information in the data is lost. It is better to employ coordinates in which the singularity trouble does not arise. Plain rectangular coordinates would solve this problem, but would introduce a nonlinear relation between measurement variables and filter variables. It was one of our basic objectives to avoid this nonlinearity.

Coordinates that can be related to measurement variables in a one-to-one fashion are simple spherical coordinates, and it is these coordinates that were decided upon. A singularity in the various functions arising in the algorithms can arise only on the polar axis of the spherical coordinates. This singularity is not only less likely to occur, but is more easily managed if it should become necessary.

2.1 System Dynamics and the Guidance Algorithm

We here repeat the equations of motion and guidance algorithm as previously developed.^{1,2}

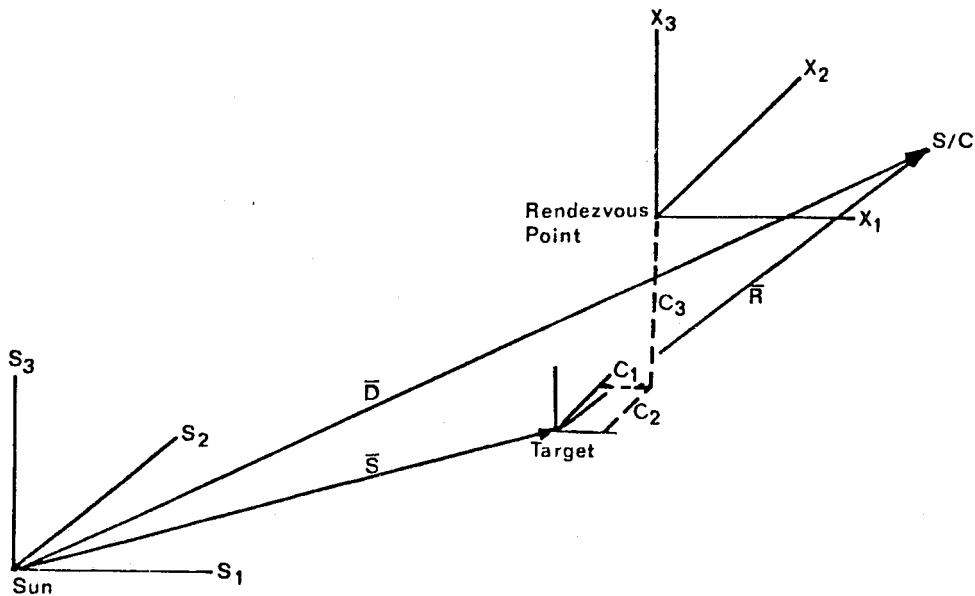


Figure 1. Rendezvous Geometry.

Appropriate to onboard guidance, the equations of motion are set in a coordinate frame fixed relative to the target as shown in Figure 1. In rectangular coordinates centered at the rendezvous point, the equations of motion, neglecting the gravitational attraction of the target, are

$$\begin{aligned}
 \dot{x}_1 &= x_4 \\
 \dot{x}_2 &= x_5 \\
 \dot{x}_3 &= x_6 \\
 \dot{x}_4 &= F_1 + \frac{GM}{S^3} [S_1 - D_1 (S/D)^3] \\
 \dot{x}_5 &= F_2 + \frac{GM}{S^3} [S_2 - D_2 (S/D)^3] \\
 \dot{x}_6 &= F_3 + \frac{GM}{S^3} [S_3 - D_3 (S/D)^3]
 \end{aligned} \tag{1}$$

where the subscripts indicate components in the corresponding coordinate directions. M is the mass of the sun and G is the universal gravitational constant. From the optimal control theory guidance law, the control forces, F_1 , F_2 , F_3 , which constitute the guidance algorithm, are

$$\begin{aligned}
 F_1 &= \left[\frac{6}{\tau_o^2} \left(1 - \frac{2\tau}{\tau_o} \right) \right] x_{10} + \left[\frac{2}{\tau_o} \left(1 - \frac{3\tau}{\tau_o} \right) \right] x_{40} \\
 F_2 &= \left[\frac{6}{\tau_o^2} \left(1 - \frac{2\tau}{\tau_o} \right) \right] x_{20} + \left[\frac{2}{\tau_o} \left(1 - \frac{3\tau}{\tau_o} \right) \right] x_{50} \\
 F_3 &= \left[\frac{6}{\tau_o^2} \left(1 - \frac{2\tau}{\tau_o} \right) \right] x_{30} + \left[\frac{2}{\tau_o} \left(1 - \frac{3\tau}{\tau_o} \right) \right] x_{60}
 \end{aligned} \tag{2}$$

where x_{10} , x_{20} , etc., are the initial conditions and $\tau = T_F - T$ is the time to go with T_F = final rendezvous time and T = running time.

Note that the equations above are in rectangular coordinates. These coordinates are employed for the precision integration made to construct the comparison trajectory for the evaluation to be made. Transformation to the spherical coordinates for the measurement and filter variables are made in the navigation algorithm.

2.2 The Navigation Algorithm in Spherical Coordinates.

As in previous work a procedure devised by Mehra³ was chosen to handle nonlinearities of the state-measurement transformation. In this procedure, the filtering is carried out in measurement variable coordinates where the observation transformation is linear.

The state of the system, $\underline{x} = (x_1 x_2 x_3 x_4 x_5 x_6)^T$, is transformed from rectangular coordinates used for the guidance problem to spherical coordinates, $\underline{y} = (R \lambda \beta \dot{R} \dot{\lambda} \dot{\beta})^T$ as defined in Figure 2. Measurements of, for example, range, range rate, and angles are a subset of the spherical coordinates. For this reason the y variables are called measurement variables.

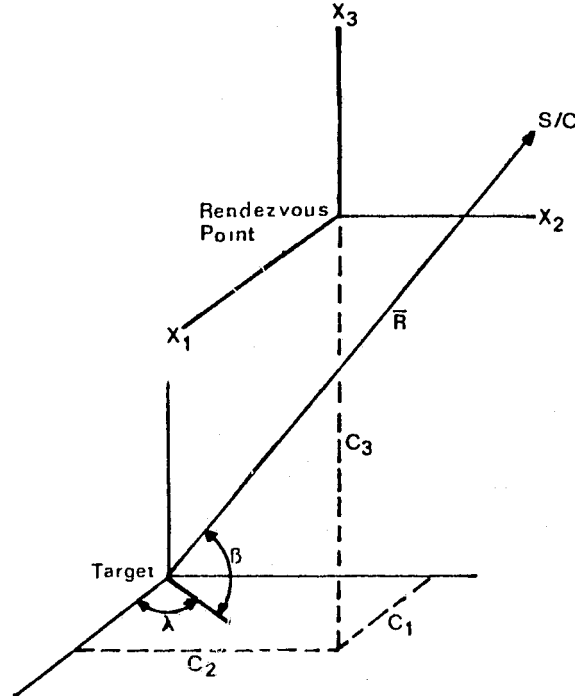


Figure 2. Transformation Geometry.

The transformation from rectangular coordinates, centered at the rendezvous point, to the measurement variables, $\underline{y} = g(\underline{x})$, is

$$\begin{aligned}
 R &= [(x_1+C_1)^2 + (x_2+C_2)^2 + (x_3+C_3)^2]^{\frac{1}{2}} \\
 \lambda &= \text{ARCTAN} [(x_2+C_2)/(x_1+C_1)] \\
 \beta &= \text{ARCTAN} \left[\frac{x_3 + C_3}{[(x_1+C_1)^2 + (x_2+C_2)^2]^{\frac{1}{2}}} \right] \\
 \dot{R} &= [(x_1+C_1)x_4 + (x_2+C_2)x_5 + (x_3+C_3)x_6]/R \\
 \dot{\lambda} &= [(x_1+C_1)x_5 - (x_2+C_2)x_4]/[(x_1+C_1)^2 + (x_2+C_2)^2] \\
 \dot{\beta} &= \frac{[(x_1+C_1)^2 + (x_2+C_2)^2]x_6 - (x_3+C_3)[(x_1+C_1)x_4 + (x_2+C_2)x_5]}{[(x_1+C_1)^2 + (x_2+C_2)^2]^{\frac{1}{2}} R^2}
 \end{aligned} \tag{3}$$

The inverse transformation, $\underline{x} = \ell(\underline{y})$ is

$$\begin{aligned}
 x_1 &= R \cos \lambda \cos \beta - C_1 \\
 x_2 &= R \cos \beta \sin \lambda - C_2 \\
 x_3 &= R \sin \beta - C_3 \\
 x_4 &= \dot{R} \cos \lambda \cos \beta - R\dot{\beta} \sin \beta \sin \lambda + R\dot{\lambda} \cos \beta \sin \lambda \\
 x_5 &= \dot{R} \cos \beta \sin \lambda - R\dot{\beta} \sin \beta \sin \lambda + R\dot{\lambda} \cos \beta \cos \lambda \\
 x_6 &= \dot{R} \sin \beta + R\dot{\beta} \cos \beta
 \end{aligned} \tag{4}$$

It was assumed that the most general set of measurements that can be made is range, angles λ and β , and range rate. It was assumed that angular rates $\dot{\lambda}$ and $\dot{\beta}$ cannot be measured directly.

The filter process proceeds as follows. Starting with an estimate $\hat{\underline{x}}_{k/k}$ at time T_k , and the spirit of the linear Kalman theory, a state estimate $\hat{\underline{x}}_{k+1/k}$ at time T_{k+1} is obtained by a linear extrapolation through the state transition matrix for the linearized system

$$\hat{\underline{x}}_{k+1/k} = \phi_{k+1/k} \hat{\underline{x}}_{k/k} \tag{5}$$

where

$$\phi_{k+1/k} = \begin{bmatrix} \alpha_{k+1/k} & 0 & 0 & \beta_{k+1/k} & 0 & 0 \\ 0 & \alpha_{k+1/k} & 0 & 0 & \beta_{k+1/k} & 0 \\ 0 & 0 & \alpha_{k+1/k} & 0 & 0 & \beta_{k+1/k} \\ \gamma_{k+1/k} & 0 & 0 & \delta_{k+1/k} & 0 & 0 \\ 0 & \gamma_{k+1/k} & 0 & 0 & \delta_{k+1/k} & 0 \\ 0 & 0 & \gamma_{k+1/k} & 0 & 0 & \delta_{k+1/k} \end{bmatrix} \quad (6)$$

and,

$$\begin{aligned} \alpha_{k+1/k} &= \frac{\tau_{k+1/k}}{\tau_{k/k}} \left[3 \left(\frac{\tau_{k+1/k}}{\tau_{k/k}} \right) - 2 \left(\frac{\tau_{k+1/k}}{\tau_{k/k}} \right)^2 \right] \\ \beta_{k+1/k} &= \tau_{k+1/k} \left[\frac{\tau_{k+1/k}}{\tau_{k/k}} - \left(\frac{\tau_{k+1/k}}{\tau_{k/k}} \right)^2 \right] \\ \gamma_{k+1/k} &= -\frac{6}{\tau_{k/k}} \left[\left(\frac{\tau_{k+1/k}}{\tau_{k/k}} \right) - \left(\frac{\tau_{k+1/k}}{\tau_{k/k}} \right)^2 \right] \\ \delta_{k+1/k} &= 3 \left(\frac{\tau_{k+1/k}}{\tau_{k/k}} \right) - 2 \left(\frac{\tau_{k+1/k}}{\tau_{k/k}} \right)^2 \end{aligned} \quad (7)$$

with time to go given by

$$\begin{aligned} \tau_{k/k} &= T_F - T_k \\ \tau_{k+1/k} &= T_F - T_{k+1} \end{aligned} \quad (8)$$

Another method used to form the first state estimate $\hat{x}_{k+1/k}$ is to directly integrate (fourth-order Runge-Kutta) the state estimate

$$\hat{x}_{k+1/k} = \hat{x}_{k/k} + \int_{T_k}^{T_{k+1}} \dot{\hat{x}} dt \quad (9)$$

With new computing equipment, this integration can be done onboard.

After the first estimate $\hat{x}_{k+1/k}$ is formed, we then transfer to measurement variables with Equations (3) in the form

$$z_{k+1/k} = g(\hat{x}_{k+1/k}) \quad (\text{nonlinear}) \quad (10)$$

The best estimate of the measurement variables at T_{k+1} is then given by Kalman's relation

$$\hat{y}_{k+1/k+1} = \hat{y}_{k+1/k} K_{k+1/k} (z_{k+1} - H\hat{y}_{k+1/k}) \quad (11)$$

where H is a rectangular matrix of ones and zeros that picks from $\hat{y}_{k+1/k}$ those elements that correspond to the actual measurements, which, in this case, is

$$H = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix} \quad (12)$$

z_{k+1} is the actual measurement vector, and $K_{k+1/k}$ is the Kalman gain matrix (yet to be calculated). After filtering, transformation is effected back to rectangular coordinates using Equations (4) in the form

$$\hat{x}_{k+1/k+1} = \mathcal{L}(y_{k+1/k+1}) \quad (\text{nonlinear}) \quad (13)$$

The Kalman gain is calculated by

$$K_{k+1/k} = M_{k+1/k} H^T (H M_{k+1/k} H^T + R_{k+1})^{-1} \quad (14)$$

where R_{k+1} is a square diagonal matrix of the variances of the measurement errors, and $M_{k+1/k}$ is a matrix consisting of the transferred covariance of measurement variables calculated by

$$M_{k+1/k} = \psi_{k+1/k} M_{k/k} \psi_{k+1/k}^T \quad (15)$$

where $M_{k/k}$ is the measurement variables covariance matrix at T_k and $\psi_{k+1/k}$ is an equivalent "transition matrix" for the measurement variables; i.e.,

$$\hat{y}_{k+1/k} \approx \psi_{k+1/k} \hat{y}_{k/k} \quad (16)$$

Mehra observed that ψ can be constructed by calculating

$$\psi_{k+1/k} = \left(\frac{\partial y_{k+1/k}}{\partial y_{k/k}} \right) \left(\frac{\partial y_{k/k}}{\partial x_{k/k}} \right) \left(\frac{\partial x_{k/k}}{\partial y_{k+1/k}} \right)$$

or

$$\psi_{k+1/k} = \left(\frac{\partial g}{\partial \underline{x}} \right)_{k+1/k} \phi_{k+1/k} \left(\frac{\partial \ell}{\partial \underline{y}} \right)_{k/k} \quad (17)$$

The matrices $(\partial \ell / \partial \underline{y})_{k/k}$ and $\phi_{k+1/k}$ are available from the best estimate of state at $T_{k/k}$, and $(\partial g / \partial \underline{x})_{k+1/k}$ is formed by using the first estimate at $T_{k+1/k}$. The matrices $(\partial \ell / \partial \underline{y})$ and $(\partial g / \partial \underline{x})$, in columns, are

$$\begin{aligned}
 \left(\frac{\partial \ell}{\partial y_1}\right) &= \begin{bmatrix} \cos \lambda \cos \beta \\ \sin \lambda \cos \beta \\ \sin \beta \\ -\dot{\beta} \cos \lambda \sin \beta - \dot{\lambda} \sin \lambda \cos \beta \\ \dot{\lambda} \cos \lambda \cos \beta - \dot{\beta} \sin \lambda \sin \beta \\ \dot{\beta} \cos \beta \end{bmatrix} \\
 \left(\frac{\partial \ell}{\partial y_2}\right) &= \begin{bmatrix} -R \sin \lambda \cos \beta \\ R \cos \lambda \cos \beta \\ 0 \\ -\dot{R} \sin \lambda \cos \beta + R \dot{\beta} \sin \lambda \sin \beta - R \dot{\lambda} \cos \lambda \cos \beta \\ -\dot{R} \cos \lambda \cos \beta - R \dot{\lambda} \sin \lambda \cos \beta - R \dot{\beta} \cos \lambda \sin \beta \\ 0 \end{bmatrix} \\
 \left(\frac{\partial \ell}{\partial y_3}\right) &= \begin{bmatrix} -R \cos \lambda \sin \beta \\ -R \sin \lambda \sin \beta \\ R \cos \beta \\ -\dot{R} \cos \lambda \sin \beta - R \dot{\beta} \cos \lambda \cos \beta + R \dot{\lambda} \sin \lambda \sin \beta \\ -\dot{R} \sin \lambda \sin \beta - R \dot{\lambda} \cos \lambda \sin \beta - R \dot{\beta} \sin \lambda \cos \beta \\ R \cos \beta - R \dot{\beta} \sin \beta \end{bmatrix} \\
 \left(\frac{\partial \ell}{\partial y_4}\right) &= \begin{bmatrix} 0 \\ 0 \\ 0 \\ \cos \lambda \cos \beta \\ \sin \lambda \cos \beta \\ \sin \beta \end{bmatrix} \\
 \left(\frac{\partial \ell}{\partial y_5}\right) &= \begin{bmatrix} 0 \\ 0 \\ 0 \\ -R \sin \lambda \cos \beta \\ R \cos \lambda \cos \beta \\ 0 \end{bmatrix} \\
 \left(\frac{\partial \ell}{\partial y_6}\right) &= \begin{bmatrix} 0 \\ 0 \\ 0 \\ -R \cos \lambda \sin \beta \\ -R \sin \lambda \sin \beta \\ R \cos \beta \end{bmatrix}
 \end{aligned} \tag{18}$$

$$\begin{aligned}
 \left(\frac{\partial g}{\partial x_1} \right) &= \begin{bmatrix} P_1/R \\ -P_2/Q^2 \\ -(P_1P_3)/(QR^2) \\ (P_4R^2-VP_1)/R^3 \\ (P_5Q^2-2WP_1)Q^4 \\ (P_1P_6-P_3P_4)/(QR^2) - (2QP_1P_6)/R^4 + (2AP_1P_3)/(QR^4) + (P_1P_3A)/(Q^3R^2) \end{bmatrix} \\
 \left(\frac{\partial g}{\partial x_2} \right) &= \begin{bmatrix} P_2/R \\ P_1/Q^2 \\ -(P_2P_3)/(QR^2) \\ (R^2P_5-VP_2)/R^3 \\ -(P_4Q^2+2WP_2)/Q^4 \\ (P_2P_6-P_3P_5)/(QR^2) - (2QP_2P_6)/R^4 + (2AP_2P_3)/(QR^4) + (P_2P_3A)/(Q^3R^2) \end{bmatrix} \\
 \left(\frac{\partial g}{\partial x_3} \right) &= \begin{bmatrix} P_3/R \\ 0 \\ Q/R^2 \\ (P_6R^2-VP_3)/R^3 \\ 0 \\ -A/(QR^2) - (2QP_3P_6)/R^4 + (2AP_3^2)/(QR^4) \end{bmatrix} \\
 \left(\frac{\partial g}{\partial x_4} \right) &= \begin{bmatrix} 0 \\ 0 \\ 0 \\ P_1/R \\ -P_2/Q^2 \\ -(P_1P_3)/(QR^2) \end{bmatrix} \\
 \left(\frac{\partial g}{\partial x_5} \right) &= \begin{bmatrix} 0 \\ 0 \\ 0 \\ P_2/R \\ P_1/Q^2 \\ -(P_2P_3)/(QR^2) \end{bmatrix} \\
 \left(\frac{\partial g}{\partial x_6} \right) &= \begin{bmatrix} 0 \\ 0 \\ 0 \\ P_3/R \\ 0 \\ Q/R^2 \end{bmatrix}
 \end{aligned}
 \tag{19}$$

where

$$\begin{aligned}
 P_1 &= x_1 + C_1 & Q &= [P_1^2 + P_2^2]^{\frac{1}{2}} \\
 P_2 &= x_2 + C_2 & R &= [P_1^2 + P_2^2 + P_3^2]^{\frac{1}{2}} \\
 P_3 &= x_3 + C_3 & A &= P_1 P_4 + P_2 P_5 \\
 P_4 &= \dot{x}_1 = x_4 & V &= P_1 P_4 + P_2 P_5 + P_3 P_6 \\
 P_5 &= \dot{x}_2 = x_5 & W &= P_1 P_5 - P_2 P_4 \\
 P_6 &= \dot{x}_3 = x_6
 \end{aligned} \tag{20}$$

All that remains is to propagate the covariance to time T_{k+1} and this is done by the relation

$$M_{k+1/k+1} = (I - K_{k+1/k} H) M_{k+1/k} \tag{21}$$

2.3 The Modified Guidance and Navigation Scheme

The procedure is illustrated in the block diagram in Figure 3. Starting at time $T_{k/k}$ an estimate of the state $\hat{x}_{k/k}$ is presumed available for evaluation. The exact state x_k is also specified at this time. The estimate $x_{k/k}$ is put into the guidance law to generate the thrusting required over the ensuing guidance interval. The full equations of motion of the comet are then integrated accurately to a time T_{k+1} and the result is then transformed to measurement variables and approximate noise added to simulate actual measurements z_{k+1} . To represent onboard computations, the state $\hat{x}_{k/k}$ is propagated to time T_{k+1} through the transition matrix $\phi_{k+1/k}$ (or integrated as shown by dotted line). The nonlinear transformation $g(x)$ to measurement variables is then made to give a first estimate $\hat{y}_{k+1/k}$. The filtered estimate $\hat{y}_{k+1/k+1}$ is then transformed nonlinearly by $\ell(y)$ to obtain the new state estimate $\hat{x}_{k+1/k+1}$ and the process is repeated.

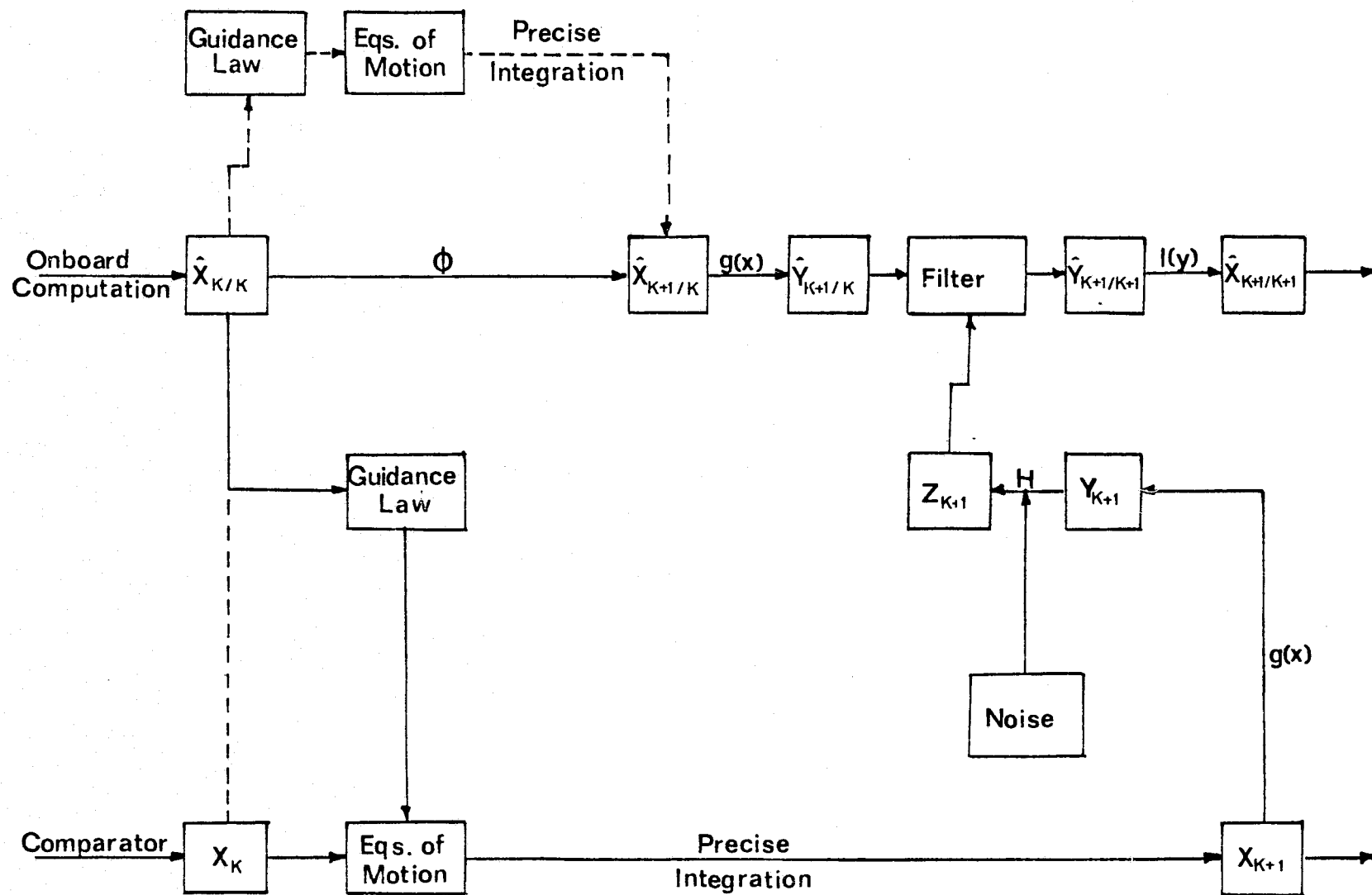


Figure 3. Evaluation Scheme.

3.0 ALGORITHM DEVELOPMENT

Having completed the reformulation in spherical coordinates, a second task is to reinvestigate the divergence problem. Proper choice of statistical quantities employed to set the filter is necessary, and work done to provide this proper choice is outlined in Section 3.1.

Other approaches to improvement of system performance through trajectory biasing and end point control are described in Sections 3.2 and 3.3.

3.1 Investigation of Filter Divergence

The Kalman filter theoretically produces an increasingly accurate estimate as additional data is processed. However, under actual operating conditions, error levels in the Kalman filter are often higher than predicted by theory. Errors can, in fact, increase continuously although additional data is being processed. No general way of handling this problem is available and approximations based on ideas from the theory and ad hoc procedures are the rule and this is the approach we take. To insure that the set of measurements employed are sufficient to reconstruct all states, system observability is investigated first.

3.1.1 System Observability

A system is observable if all states can be reconstructed from the set of available measurements. For the equations

$$\begin{aligned} \underline{y}_{k+1/k} &= \psi_{k+1/k} \underline{y}_{k/k} \\ \underline{z}_{k+1} &= H \underline{y}_{k+1} \end{aligned} \tag{22}$$

an augmented matrix derived by Kalman⁴

$$F = (H^T, \psi^T H^T, (\psi^T)^2 H^T, \dots, (\psi^T)^{n-1} H^T) \quad (23)$$

is formed. If the rank of the matrix F is n , where n is the order of the y vector, the system is observable.

A computer program written by Bullock and Fosha⁵ is used to form the matrix F . The matrix is normalized at each step $(H^T, \psi^T H^T, \dots)$ so that each column has unit length. The complete matrix is renormalized so each row has unit length (The normalization procedure prevents overflow in large problems). Then the square matrix FF^T is computed. If FF^T is nonsingular, the system is observable.

Since ψ is a function of time, the matrix F is also a function of time. Therefore, F must be computed at each guidance interval (or any multiple) because a new F is present at each step, and the degree of observability can change.

3.1.2 State Covariance Weighting Criteria

In nonlinear systems, the state error covariance matrix tends to reduce rapidly with respect to the actual state indeterminacies. This can be attributed to the fact that the filter believes the first measurement is highly accurate when it is not, and state estimates are biased incorrectly. To account for the reduction of the covariance matrix, a weighting procedure is employed.

Introducing state disturbance into the system, equation (16) becomes

$$y_{k+1/k} = \psi_{k+1/k} y_{k/k} + \Gamma w_k \quad (24)$$

where Γ is the disturbance transition matrix, assumed to be identity, and \underline{w}_k is the disturbance vector. The process $\{\underline{w}_k, k=0,1,\dots\}$ is a Gaussian White sequence for which

$$E[\underline{w}_k] = 0, \quad h = 1, 2, \dots \quad (25)$$

Defining the positive definite matrix

$$Q_k = E[\underline{w}_k \underline{w}_k^T] \quad (26)$$

It can be shown (See, for example, Meditch⁶) that the state error covariance matrix is given by the relation

$$M_{k+1/k} = \psi_{k+1/k} M_{k/k} \psi_{k+1/k}^T + Q_k \quad (27)$$

This weighting criteria aids in eliminating the severe reduction of the state covariance matrix and keeps it from becoming non-positive definite. However, proper weighting of measurements is not guaranteed because Q cannot be accurately determined, and errors can still exist.

3.1.3 Approximation of the Initial State Error Covariance Matrix

In a nonlinear system, good state knowledge at time k does not insure good knowledge of future state because of nonlinearities in the dynamical equations. Hence, the Kalman gain must weight the measurements sufficiently to compensate for the errors in the one-step propagation. Measurements are not weighted enough initially when the filter is initialized with state error covariance matrices that are too small. To alleviate this problem, a design procedure for determining a suitable initial state error covariance matrix is introduced.

Assuming a priori information for the range of values of the Kalman gain, a suitable initial state error covariance can be found eliminating the incorrect weighting of measurements. The only elements that are

directly adjustable in the Kalman gain matrix are those which lie on the diagonal of the 6x4 gain matrix (to be shown later; these adjustable elements correspond directly to the first four diagonal elements of the state covariance matrix because of the special form of the observation matrix (discussed previously). Therefore, for given values of the Kalman gain, an initial state covariance matrix (four diagonal elements) can be found.

There exists a one-to-one relationship between the diagonal elements of the Kalman gain matrix and the first four diagonal elements of the state covariance matrix. A change in one element of the state covariance matrix results in a direct change of the corresponding element in the Kalman gain. This can be shown by performing the matrix operations. By precalculation, ψ is found to be approximately an identity matrix, so Equation (27) can be written

$$M_{k+1/k} \approx M_{k/k} + Q_k \quad (28)$$

where $M_{k+1/k}$ is

$$M_{k+1/k} = \begin{bmatrix} M_{11} & & & & & \\ & M_{22} & & & & \\ & & M_{33} & & & \\ & & & M_{44} & & \\ & & & & M_{55} & \\ & & & & & M_{66} \end{bmatrix}$$

The Kalman gain relationship is given by

$$K_{k+1/k} = M_{k+1/k} H^T (H M_{k+1/k} H^T + R_{k+1})^{-1} \quad (29)$$

We write out Equation (29) in explicit matrix form,

$$\begin{bmatrix} K_{11} & & & & & \\ & K_{22} & & & & \\ & & K_{33} & & & \\ & & & K_{44} & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \end{bmatrix}_{6 \times 4} = \begin{bmatrix} M_{11} & & & & & \\ & M_{22} & & & & \\ & & M_{33} & & & \\ & & & M_{44} & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \end{bmatrix}_{6 \times 4} \begin{bmatrix} M_{11}+R_{11} & & & & & \\ & M_{22}+R_{22} & & & & \\ & & M_{33}+R_{33} & & & \\ & & & M_{44}+R_{44} & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \end{bmatrix}_{4 \times 4}^{-1} \quad (30)$$

Notice that only the first four diagonal elements of the state covariance matrix have any direct effect on the Kalman gain matrix because the elements M_{55} and M_{66} were eliminated in the calculations.

The Kalman gain matrix can be reduced to a 4x4 matrix since only the diagonal elements are considered. The diagonal elements of the 4x6 observation matrix correspond directly with the adjustable elements in the gain matrix, so H is taken to be a 4x4 identity matrix. The state covariance matrix is also reduced to a 4x4 matrix since only four elements can be directly obtained.

Now, an initial covariance matrix can be found insuring accurate weights for the measurements. Rewriting Equation (29), the Kalman gain is

$$K_{k+1/k} = M_{k+1/k} (M_{k+1/k} + R_{k+1})^{-1} \quad (31)$$

Elementary matrix manipulations of Equation (31) result in

$$M_{k+1/k} = (I - K_{k+1/k})^{-1} K_{k+1/k} R_{k+1} \quad (32)$$

Since ψ is approximately an identity matrix, the initial state covariance matrix can be calculated by

$$M_{k/k} \approx M_{k+1/k} - Q_k \quad (33)$$

The formulation calculates only those diagonal elements in the state covariance matrix which correspond to the actual measurements. But this is sufficient because the other diagonal elements in the state covariance matrix do not affect the diagonal elements in the Kalman gain matrix. However, the other diagonal elements do affect the off-diagonal terms in the $\psi M_{k/k} \psi^T$ calculation which, in turn, affects the fifth and sixth rows of the Kalman gain matrix. Therefore, the other diagonal elements must be chosen carefully. But this can only be done through numerical experiments.

With this formulation, a range of values for the state covariance matrix corresponding to the values of the Kalman gain matrix can be calculated, and, through simulation, the desired state covariance matrix can be determined. However, since the unmeasured states are not directly weighted, divergence can still occur.

3.2 Trajectory Biasing

In the comet rendezvous problem, the optimal control algorithm tends to fly the spacecraft directly toward the desired rendezvous point. If the rendezvous point lies directly between the comet and spacecraft, as illustrated in Figure 4, the measured angles change very little in the initial stages of flight and can cause poor estimation of the angular

rates (unmeasured states). Poor estimates of the angular rates can cause the velocity of the spacecraft to be in error and accurate rendezvous cannot be attained. To eliminate this problem, trajectory biasing is investigated.

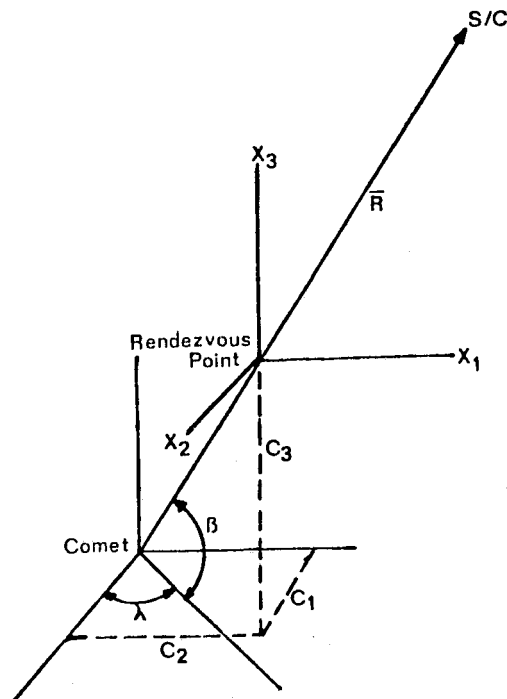


Figure 4. Possible Rendezvous Geometry.

Instead of flying directly to the desired rendezvous point, the spacecraft initially flies toward a point biased away from the target point. This point can be changed at each guidance step (or some multiple) until the biased point becomes the desired rendezvous point. Biasing the trajectory of the spacecraft in this manner gives larger angular change and, consequently, better estimation of angular rates.

With a better estimate of the unmeasured states, velocity estimates are more accurate, and more accurate rendezvous can be attained.'

3.3 End point Control

The guidance algorithm singularity leads to a "blow up" of the procedure at rendezvous unless appropriate steps are taken. One approach would be to "freeze" the commands near rendezvous, but this is not a good way because after the rendezvous is accomplished, it must be maintained. That is, station keeping must be done. For this reason we looked for a modification that would transform the terminal approach algorithm into a station keeping algorithm while avoiding the singularity. One easy way to do this is simply to push time to go a predetermined amount as rendezvous is approached. A successive increase of time to go by a predetermined amount of time at each guidance step after a specified point will keep time to go large enough to avoid the singularity and automatically turns the approach algorithm into a station keeping algorithm. For deterministic runs and evaluations with very high precision data this worked well when implemented in the form of an increase of one guidance time interval at each succeeding step starting at the time of one interval before rendezvous. However, for cases of only moderately accurate data, this procedure still led to thrusts above those easily attainable with SEP (about $10^{-4}g$). Several other methods of reducing end point thrust were considered, and a particularly simple one was found to work well. Instead of pushing one step, we chose an increase in time to go equal to

$$\left| \frac{F}{10^{-4}} \right| \Delta$$

Where,

Δ = guidance step size

10^{-4} = maximum SEP thrust (g's)

F = value of thrust in unmodified
next interval.

Here, the symbol $|A|$ means round A to the next higher integer. Note that there are extensions of this technique that could be applied early in the terminal approach should such be required. However, this extension was not incorporated in the computer program as it was not necessary for the cases considered.

4.0 ALGORITHM EVALUATION

The full guidance and navigation algorithm of Section 2.0 was programmed for simulation on an IBM 370/150 digital computer. The simulation package includes a double precision, fourth-order Runge-Kutta integrator to provide an accurate comparison trajectory for evaluation of algorithm performance. A brief discussion of the program and user instructions are included in the Appendix.

4.1 Development Simulations

The starting point was chosen to be five days before rendezvous at a distance of 50,000 km from the target and a relative velocity of 20,000 km/day. Measurements are made and navigation computations made each one-tenth of a day. The rendezvous point is located 1732 km from the target with 1000 km displacement in each coordinate direction. In the first simulations, the initial state error covariance was assumed to be a diagonal matrix with terms of the order of 10^5 in units of km and days. This matrix is an identity matrix in the units of 10^5 km for distance and days for time as used in the computer program. A five percent error in each coordinate of the true initial state was chosen as the estimated position and velocity of spacecraft.

Simulations were made and gross filter divergence was found as shown in Figure 5. Terminal rendezvous errors were about 2000 km and 1,500 km/day (17.4 m/sec). Divergence began approximately four days from rendezvous. At this point, system observability was checked with the method of Section 3.1.1 to determine if the four measurements were

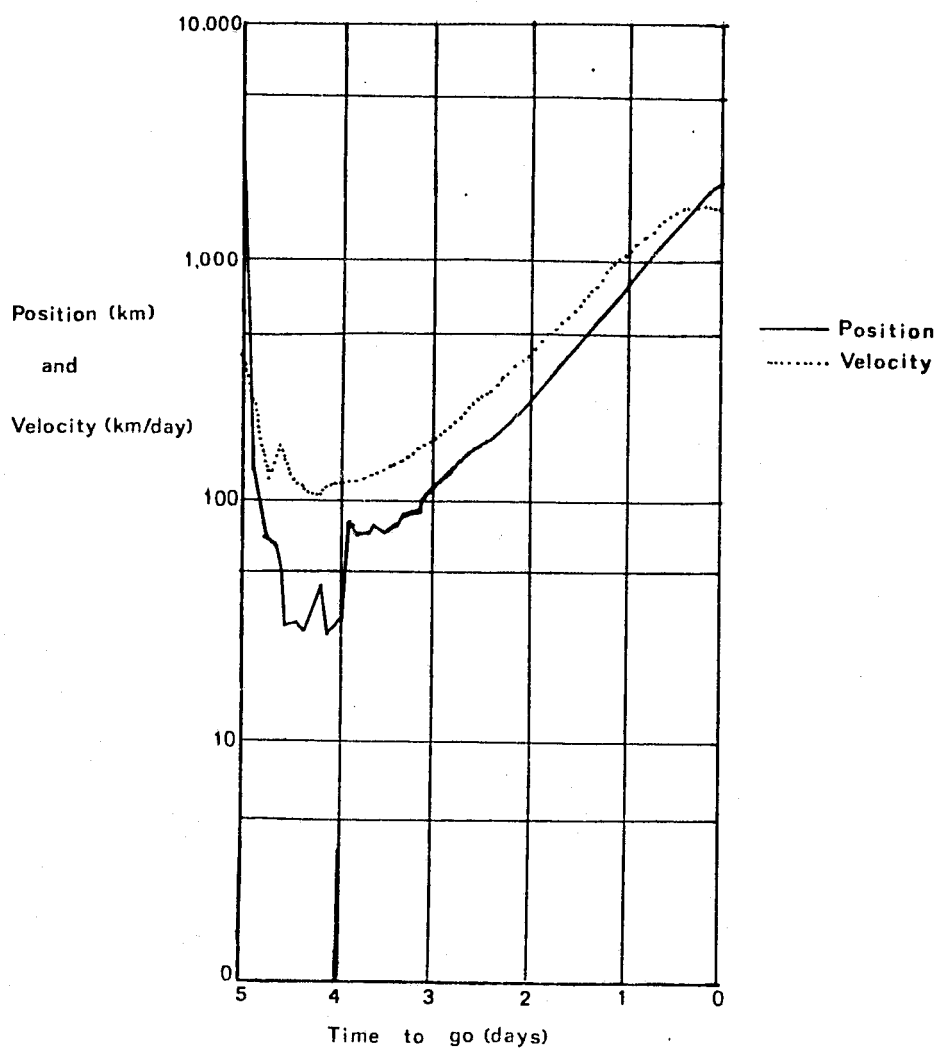


Figure 5. Navigation Errors for Initial Simulations.

sufficient to reconstruct all six states. The observability criteria was applied with solutions showing complete system observability throughout the mission. Hence, methods for correcting divergence were investigated.

One cause of divergence was the rapid reduction of the state error covariance matrix. In fact, simulations show the covariance matrix became numerically zero in the computer near rendezvous and completely ignored new measurements as they were made. A standard fix is to insert noise into the basic system equations as discussed in Section 3.1.2. A constant matrix Q was added to the state error covariance matrix to approximate this noise. Q was arbitrarily chosen diagonal with elements of the order of magnitude of 10. The magnitude of the diagonal elements in Q are approximately the same size as those in the state error covariance matrix when divergence started.

Figure 6 shows a marked improvement in the navigation errors with the addition of the Q matrix. Errors at rendezvous were approximately 11.5 km and 310 km/day (3.59 m/sec) with the spacecraft a distance of 28 km from rendezvous.

Continuing the divergence investigation the method of Section 3.1.3 was employed to estimate initial covariance. Assuming desired values for the diagonal elements of the Kalman gain matrix of about .95, an initial state error covariance matrix was found. Interestingly, results were very close to the previously assumed identity matrix and we therefore continued to use the identity matrix.

The principal errors were in the velocity estimates. To get a better angular data; that is, data with larger changes, the trajectory of the spacecraft was biased away from the final target point following the idea of Section 3.2. The new target point was chosen as a function of position by the equation

$$C_1 = \{[\hat{x}_1^2 + \hat{x}_2^2 + \hat{x}_3^2]^{1/2}/1.5\} + 1000$$

where C_1 is the standoff distance in the x_1 direction. The point changes each guidance step, and, upon a forced command, becomes the

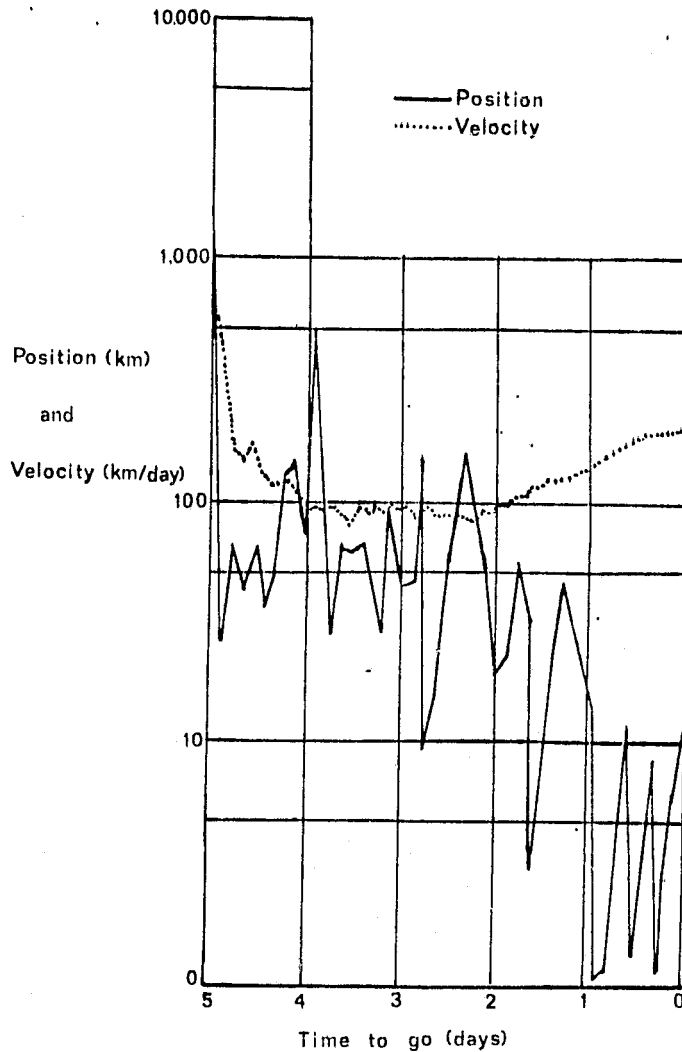


Figure 6. Navigation Errors with Q Matrix Added.

desired rendezvous point at time zero. Biasing in this manner allows the angles to change more in the early stages of flight and thus gives a more accurate estimate of the angular rates. Figure 7 shows the

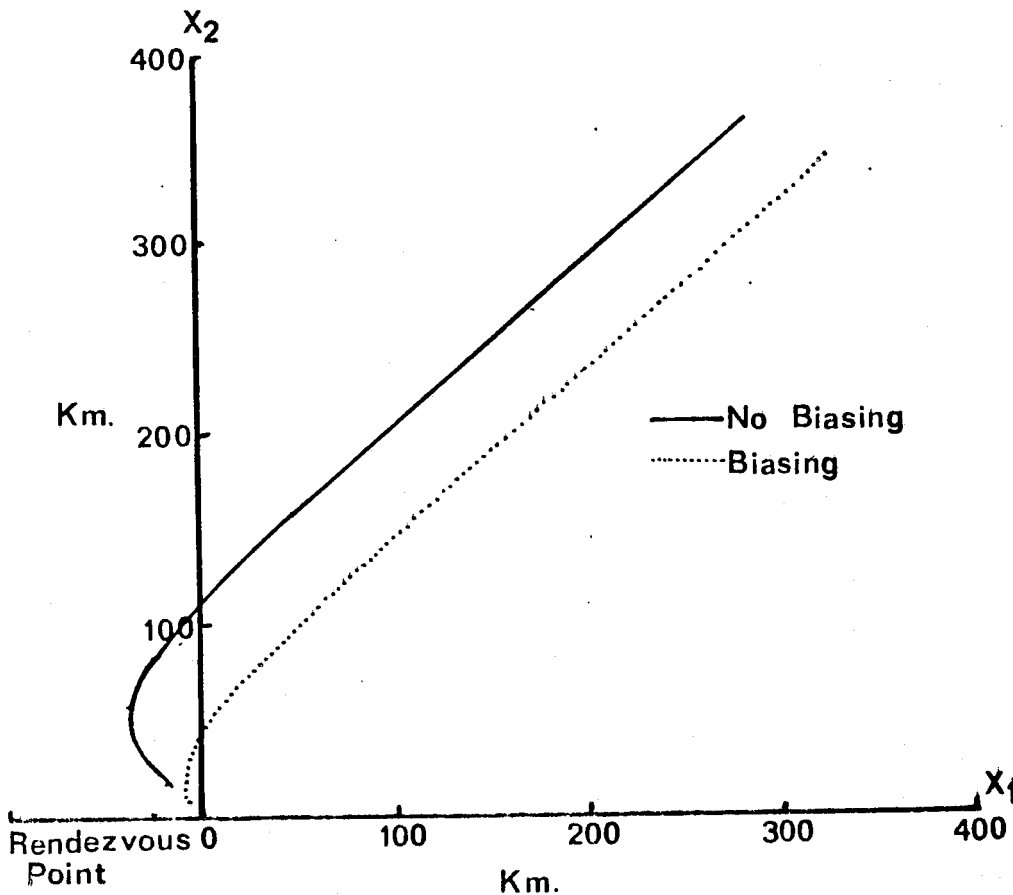


Figure 7. Comparison of Trajectories with and without Biasing within 1/2 Day to Go.

terminal geometry of typical biased and unbiased trajectories. A considerable improvement in accuracy was obtained. The improvement was sufficient to warrant undertaking the parametric study of the effects of measurement accuracy. Before conducting this study, the end point control scheme of Section 3.3 was incorporated into the program.

4.2 Generation of Parametric Data

Table 1 presents nine of the data error sets employed in the parametric study. The choice of error in range measurement standard deviation, σ_R , of 0.3 percent and 3.0 percent of range was made to bracket expected errors with realistic radar equipment. Specific information on what these errors might be is not available and these values were chosen after discussions with NASA radar electronics personnel. The choice of standard deviation of range rate measurements, $\sigma_{\dot{R}}$, of 100 km/day and 1,000 km/day (1.16 m/sec and 11.6 m/sec), was based on these same discussions. The standard deviation of 1.0 arc seconds on angle measurements was also a best estimate of what could be done with on board equipment. This value is three times larger than the value used by JPL for use of ground based data reduction of spacecraft TV.

TABLE 1. DATA SETS FOR PARAMETRIC STUDY
(Note: 1,000 km/day \approx 11.6 m/sec.)

Data Set	Range, R	Angles λ & β (arc min.)	Range Rate, \dot{R} (km/day)
A	.03156 R	1.08	10
B	↓	↓	100
C	.03 R	1.0	100
D	↓	↓	1,000
E	↓	↓	10,000
F	↓	↓	100,000
G	.003	↓	10
H	↓	↓	100
J	↓	↓	1,000

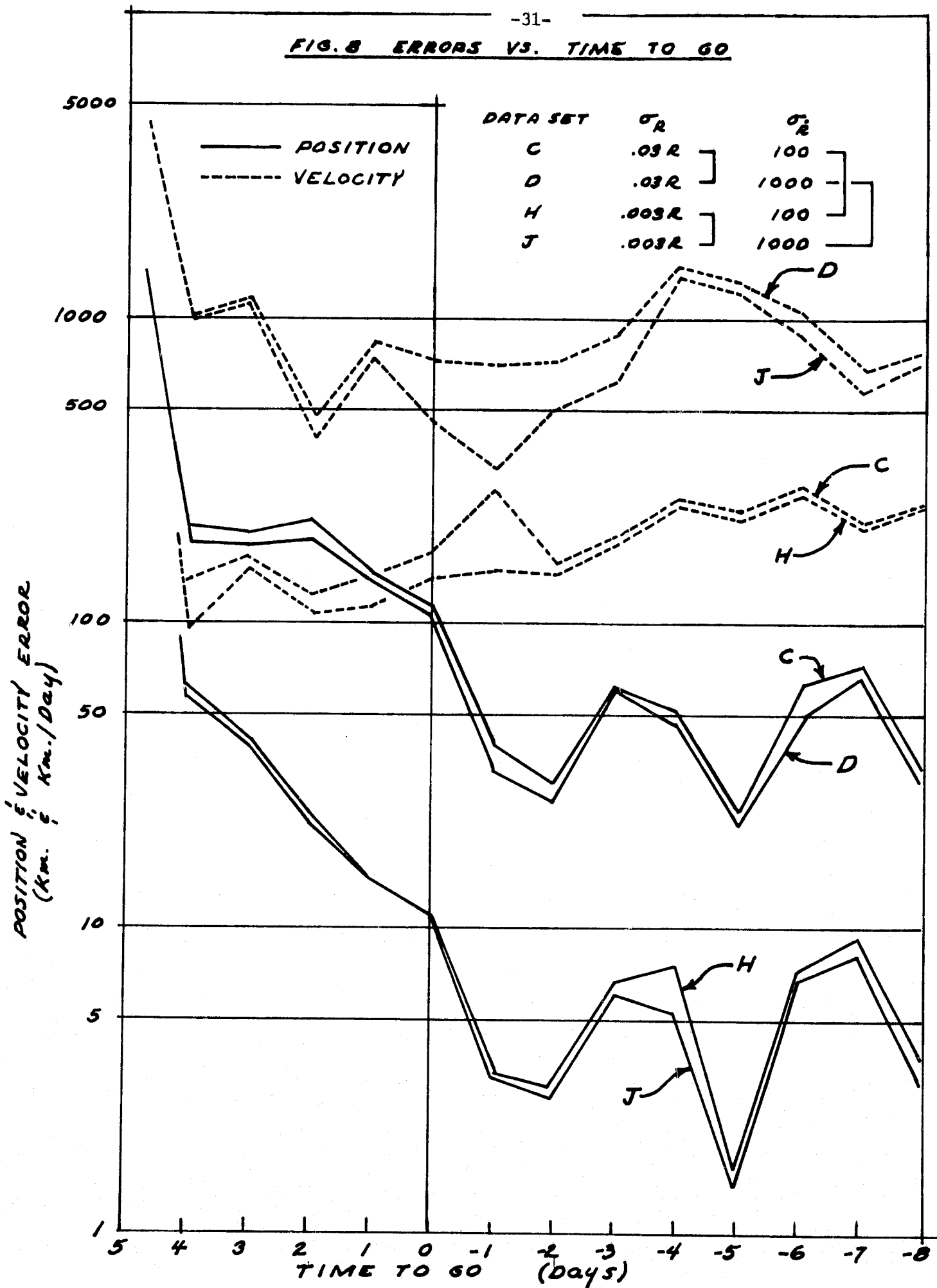
5.0 RESULTS

The parametric runs made with the data sets of Table 1 gave no cases of divergence even with the extreme range rate errors of sets E and F. And station was maintained after rendezvous for as long a period as run (up to 30 days after rendezvous). The results indicate that the rate measurements may not be required, but this requires additional investigation.

Results of four representative sets, C, D, H and J are shown in Figure 8. These sets bracket expected measurement errors as discussed in Section 4.2. In the samples, the same four sequences of random numbers was used to simulate the four measurements made (range, range-rate, and two angles) Other sequences gave results differing only in detail not accuracy levels. The plots give position and velocity errors only at intervals of one day because plots with the one-tenth day computation steps used in the simulations are cluttered and difficult to interpret. The values at each day were obtained by averaging over an interval about each plotted point.

One point is immediately obvious from Figure 8: The magnitude of the error in position depends almost completely on the accuracy of range measurements and the magnitude of the error in velocity depends almost completely on the accuracy of range rate measurement. For example, consider sets C and D. These sets have the same range measurement standard deviation (.03 R), but different rate standard deviation (100 km/day and 1000 km/day). The position errors for these two cases are very nearly the same, but the velocity errors differ by a factor

FIG. 8 ERRORS VS. TIME TO GO



of three to four. Similarly, sets C and H show essentially the same velocity errors and a difference in position error by a factor of nine or ten.

We are led to two conclusions applicable in the range of measurement errors considered: First, that reduction of range rate error by a factor of ten leads to an improvement in velocity estimation by a factor of three or four and has little effect on accuracy of position estimation. Second, that reduction in range error by a factor of ten leads to an improvement in position estimation by a factor of nine or ten and has little effect on accuracy of velocity estimation. The plots in Figure 8 show these effects add linearity. Stated another way: An improvement in range rate measurement accuracy produces only about one-third the improvement in velocity estimate while an improvement in range measurement accuracy leads to about the same improvement in position estimate. Note that the linearity of the relation for velocity breaks down and (fortunately) very large rate measurement errors do not lead to large velocity estimate errors. Range is clearly the more valuable data type.

Examination of the velocity error curves of Figure 8 shows what might seem to be a slight divergence after the nominal rendezvous time. However, the reduction in accuracy is caused by the elimination of trajectory biasing shortly before zero time to go. The ensuing motion is a cyclic motion near the rendezvous point. This is reflected in the position error curves. An appropriate biasing of the trajectory and thrust levels can undoubtedly smooth this out. This would be a point

for study only later in guidance and navigation scheme development. As it is, the errors are not large and do not grow with time.

The work presented here has lost much of its sense of immediacy because of changes in the overall NASA space program. Missions to comets or other deep space bodies which will require autonomous navigation and guidance are a long way in the future. When the time comes, the results here may help to give a starting point for the detailed investigations and development that will then be required.

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APPENDIX
COMPUTER SIMULATION PROGRAM

The computer program used is written in FORTRAN IV and used with the IBM 370/155. The program is a research tool, not a production routine. The steps in the simulation and the names of the subroutines that carry out these steps are as follows.

A fourth order Runge-Kutta subroutine, RUNKUT, is used to integrate the dynamic forces in GOFX\$ over subintervals of length DT. At each time step (DELT), observations are made in OBSERV and the filter is used to predict the state in FILTER. The information generated is transferred to subroutine CYCLOT and terminal conditions are checked. Program sequencing and execution is controlled by subroutine CYCLE. Subroutine TARGET is used to generate the comet's position. NOISE is a dummy name for the functions URAND (Uniformly Distributed Random Numbers) and GRAND (Gaussian Distributed Random Numbers). Ten independent noise channels are shared by these two functions.

A. Subroutine Names and Descriptions

MAIN	Reads in system data and calls CYCLE.
CYCLE	Controls sequence of operation and transfer of data between XT (true state) and XP (predicted state).
RUNKUT	Fourth order Runge-Kutta integrator. Dynamics are provided by DOFX\$ and guidance by GOFX\$. Called by CYCLE. Performs N integrations of step size DT at each call.

Entries:

RKINIT called by CYCLE

Initialize internal variables and read in XT

DOFX\$ - Compute contribution of dynamics to true $\dot{\underline{x}}$. J is the index of the components of XT.

Entries:

DOFX\$: Compute data to be used by all components.

DOFX: Compute each component of the true state vector.

DXINIT: Initialize internal constants and read in comet data.

SPECIAL: Calls target for comet position.

GOEFX\$: - Same as DOEF\$ except XP is used as variable.

FILTER - User supplied algorithm to calculate XP. (Basic Extended Kalman filter used in present listing).

Entries:

FLINIT: Used to initialize arrays.

SPECIAL: Calls MINV, CONOBS, and can call INTGR.

OBSERV - Generates Z; may call GRAND.

CYCLOT - Outputs data and checks for end of run.

Entries:

TERMIN: Check for end conditions to be satisfied.

RECAP: If end conditions met outputs minimum normed distance, velocity, and associated times.

CYINIT: Initialize internal constants.

TARGET - Comet's position by solution of Kepler's equation.

MINV - Gaussian elimination inversion routine.

CONOBS - Determines degree of observability

SPECIAL - Calls MTRANS, MMUL, MXINV, and NORML

MATRIX Subroutine name for group of entries.

Entries:

MMUL: Performs a special matrix multiplication for CONOBS.

MTRANS: Transposes matrices for CONOBS.

MXINV: Calls DINVER

NORML: Normalizes matrices for CONOBS.

DINVER - Matrix inversion routine using pivotal condensation for determination of determinants used in CONOBS.

INTGR - Same as RUNKUT except uses XP instead of XT.

GRAND - Generates Gaussian distributed random noise with given mean (RMEAN) and standard deviation (STDDEV).

LSCNT - Noise channel number. Calls URAND.

URAND - Generates random numbers over the interval [0,1].

BLOCK DATA - Initializes seed numbers for URAND.

B. Variable Names and Definitions

XT - True state vector.

XP - Predicted state vector (loaded in GXINIT).

XE - Error in state.

2 - True observations.

2P - Predicted observations.

2E - Error in observations (residuals)

L\$L - One BYTE logical array used to control sequencing of simulator.

L\$E - One BYTE Logical array for use by error monitor (not implemented).

DT - Integration stepsize (true position).

DELT - Guidance update stepsize (predicted position).
 N - DELT/DT (an integer).
 TI - Integration start time.
 TF - Integration end time.
 ISLEN - Number of elements in state vector.
 IOLEN - Number of elements in the observation vector.
 C - Rendezvous stand-off distance.
 A - Target semi-major axis.
 RN - Target mean motion.
 EPS - Target eccentricity.
 EO - Target eccentric anomaly.
 TSTAR - Guidance initiation time.

C. Input Data

<u>Card Number</u>	<u>Contents</u>	<u>Format</u>
1	Title Card	SA8
2	Run time logical flags (L&L)	80L1
3	Run time error flags (L&E)	80L1
4	N, ISLEN, IOLEN	8I3
5	TI, TF, DT, DELT	8F10.0
6	XT (Initial conditions)	8F10.0
7	A, RN, EPS, EO, TSTAR	8F10.0
8	C	8F10.0

D. Subroutine Initialization Entries

<u>Subroutines and Line Numbers</u>	<u>Entry</u>
GX INIT(J)	
38,39,40	Initial position error
41,42,43	Initial position error
FLINIT	
238-243	Initial state error covariance matrix
OVINIT	
90	Standard deviation of angular measurement error (EANG)
91, 92	Fix logical if statements for RMIN and RDMIN
OBSERV	
30-33	Mean, standard deviation and noise channel.
24-27	Range, angles, and range rate errors (RFRAF)

PROGRAM LISTING

MAIN

```

IMPLICIT REAL*8 (A-H,O-Z)
LOGICAL*1 L$L,L$E,LMON,LF,LT
COMMON/V$RBLE/XT(6),XP(6),XE(6),Z(6),ZP(6),ZE(6)
COMMON/T$MER/DELT,DT,TIME,TI,TF,N,ISLEN,IOLN
COMMON/SYSTEM/L$L(40),L$E(10)
COMMON/M$NITR/LMON(20)
COMMON/NOISE/IRAN(10),DG(10),RFRAF(6)
COMMON/OFFSET/C(3)
COMMON/MOQCH/T6(4,4)
DIMENSION B(4,6),RN(4,4)
DATA LF,LT/F,T/
502 FORMAT(80L1)
503 FORMAT(5A8)
504 FORMAT(8I3)
505 FORMAT(8F10.0)
602 FORMAT(1H0,80L1)
603 FORMAT(1H0,5A8)
604 FORMAT(1H0,' INPUT CARD LIST')
606 FORMAT(1H ,////)
607 FORMAT(1H0,8I5)
608 FORMAT(1H0,1P6D12.5)
WRITE(6,604)
READ(5,503)TITLE
WRITE(6,603)TITLE
READ(5,502)L$L
WRITE(6,602)L$L
READ(5,502)L$E
WRITE(6,602)L$E
READ(5,504)N,ISLEN,IOLN
WRITE(6,607)N,ISLEN,IOLN
READ(5,505)TI,TF,DT,DELT
WRITE(6,608)TI,TF,DT,DELT
WRITE(6,606)
CALL CYCLE
STOP
END
SUBROUTINE CYCLE
IMPLICIT REAL*8 (A-H,O-Z)
LOGICAL*1 L$L,L$E,LMON,LF,LT
COMMON/V$RBLE/XT(6),XP(6),XE(6),Z(6),ZP(6),ZE(6)
COMMON/T$MER/DELT,DT,TIME,TI,TF,N,ISLEN,IOLN
COMMON/SYSTEM/L$L(40),L$E(10)
COMMON/M$NITR/LMON(20)
COMMON/NOISE/IRAN(10),DG(10),RFRAF(6)
COMMON/OFFSET/C(3)
COMMON/MOQCH/T6(4,4)
DATA LF,LT/F,T/
DIMENSION B(4,6),RN(4,4)
C  **** INITIALIZE SUBROUTINES ****
CALL RKINIT
DUM=DXINIT(1)

```

```

DUM=GXINIT(1)
DUM=URINIT(1)
CALL CYINIT
CALL FLINIT
CALL OVINIT
CALL INTINI
DUM=DXININ(1)
DUM=GXININ(1)
C  **** COMPUTE TRAJECTORIES ****
1 CALL RUNKUT
  IF(L$L(2))GO TO 2
  CALL OBSERV
  CALL FILTER
  GO TO 3
2 DO 4 I=1,ISLEN
4 XP(I)=XT(I)
C  **** OUTPUT CYCLE DATA ****
3 CALL CYCLOT
C  **** MONITOR SECTION ****
  CALL TERMIN
  IF(L$L(2))GO TO 5
  IF(L$L(3))GO TO 5
  IF(L$L(6))GO TO 1
  GO TO 7
5 DO 6 I=1,ISLEN
6 XP(I)=XT(I)
  IF(L$L(6))GO TO 1
C  **** OUTPUT SECTION ****
7 CALL RECAP
  RETURN
  END
  SUBROUTINE RUNKUT
  IMPLICIT REAL*8 (A-H,O-Z)
  LOGICAL*1 L$L,L$E,L$MON,LF,LT
  LOGICAL*1 LS1,LS2
  COMMON/V$RBLE/XT(6),XP(6),XE(6),Z(6),ZP(6),ZE(6)
  COMMON/T$MER/DELT,DT,TIME,TI,TF,N,ISLEN,I$LEN
  COMMON/SY$TEM/L$L(40),L$E(10)
  COMMON/M$NITR/L$MON(20)
  COMMON/NOI$E/IRAN(10),DG(10),RFRAF(6)
  COMMON/OFFSET/C(3)
  DIMENSION XINT(6),SUM(6)
  DATA LF,LT/F,T/
  L$L(12)=LT
  YS=DSQRT(XP(1)**2+XP(2)**2+XP(3)**2)
  C(1)=1.0D-02+(YS/1.5D0)
  IF(TIME.GT.38.4) C(1)=1.0D-02
  WRITE(6,750)C
750 FORMAT(1H0,7X,'C1',11X,'C2',11X,'C3'/4X,1P3D12.5/)
  DO 1 ICYCLE=1,N
  DO 33 I=1,ISLEN

```

```

33 SUM(I)=0.000
   L$ L(10)=LT
   DO 10 I1=1,4
   LS1=I1.EQ.2. OR.I1.EQ.3
   LS2=I1.EQ.4
   L$ L(11)=I1.EQ.3
   F=F1
   FS=F5
   IF(LS1)F=F2
   IF(LS1)FS=F3
   IF(LS2)FS=F4
   TS=TIME+DT*FS
20 DO 20 I=1,ISLEN
   XINT(I)=XT(I)+FS*XINT(I)
   DO 31 I=1,NP1
   J=I-1
   IF(J.GT.0) GO TO 2
   DUM=DOFX$(J,TS)
   DUM=GOFX$(J,TS)
   L$ L(12)=LF
   GO TO 31
2   XINT(J)=DT*(DOFX(J)+GOFX(J))
   SUM(J)=SUM(J)+F*XINT(J)
31  CONTINUE
   L$ L(10)=LF
10  CONTINUE
   TIME=TIME+DT
   DO 11 I=1,ISLEN
11  XT(I)=XT(I)+SUM(I)
1   CONTINUE
   RETURN
   ENTRY RKINIT
   READ(5,501)(XT(I),I=1,ISLEN)
   WRITE(6,601)(XT(I),I=1,ISLEN)
501 FORMAT(8F10.0)
601 FORMAT(1H0,1P8D12.5)
   F1=1.000/6.000
   F2=2.000*F1
   F3=1.000/2.000
   F4=1.000
   F5=0.000
   TIME=TI
   NP1=ISLEN+1
   DO 32 I=1,ISLEN
32  XINT(I)=XT(I)
   RETURN
   END
   FUNCTION DOFX$(J,TS)
   IMPLICIT REAL*8(A-H,O-Z)
   LOGICAL*1 L$L,L$E,L$MON,LF,LT
   COMMON/V$RBLE/XT(6),XP(6),XE(6),Z(6),ZP(6),ZE(6)

```



```
COMMON/T$MER/DELT,DT,TIME,TI,TF,N,ISLEN,IOLN
COMMON/SY$TEM/L$L(40),L$E(10)
COMMON/M$NITR/LMON(20)
COMMON/NOI$E/IRAN(10),DG(10),RFRAF(6)
COMMON/VAR1/A,RN,EPS,EO,TSTAR,DET
COMMON/OFFSET/C(3)
DIMENSION D(3),S(3)
DATA LF,LT/F,T/
99601 FORMAT(1H0,I4,'IMPROPER INDEX *DOFX')
      IF(.NOT.L$L(11))CALL TARGET(S,TS)
      S2=0.000
      D2=0.000
      DO 1 I=1,3
      D(I)=S(I)+C(I)+XT(I)
      D2=D2+D(I)**2
1      S2=S2+S(I)**2
      DN=DSQRT(D2)
      SN=DSQRT(S2)
      RATI=GM/(SN*SN*SN)
      RAT2=(SN/DN)**3
      DOFX$=0.000
      IF(TF-TIME.GT.DELT)RETURN
      IF(.NOT.L$L(10))RETURN
      WRITE(6,602)TIME,XT
602  FORMAT(1H0,'END STATE ',2X,'TIME=',F10.3/1H ,1P6D12.5)
      RETURN
      ENTRY DOFX(J)
      GO TO (99999,99999,99999,99998,99998,99998),J
      WRITE(6,99601)J
      DOFX=0.000
      L$E(2)=LT
      RETURN
99999 DOFX=XT(J+3)
      RETURN
99998 DOFX=RATI*(S(J-3)-D(J-3)*RAT2)
      RETURN
      ENTRY DXINIT(J)
      READ(5,501)A,RN,EPS,EO,TSTAR
      EO=RN*TSTAR
      WRITE(6,601)A,RN,EPS,EO,TSTAR
501  FORMAT(8F10.0)
601  FORMAT(1H0,1P8D12.5)
      READ(5,501)C
      WRITE(6,601)C
      GM=9.90549D05
      DET=1.0D-3
      DXINIT=0.000
      RETURN
      END
      FUNCTION GOFX$(J,TS)
      IMPLICIT REAL*8(A-H,O-Z)
```

```

LOGICAL*1 L$L,L$E,L$M,L$F,L$T
COMMON/V$RBLE/XT(6),XP(6),XE(6),Z(6),ZP(6),ZE(6)
COMMON/T$MER/DELT,DT,TIME,TI,TF,N,ISLEN,IDLEN
COMMON/SYSTEM/L$L(40),L$E(10)
COMMON/M$NITR/L$M(20)
COMMON/NOISE/IRAN(10),DG(10),RFRAF(6)
COMMON/FORCE/F(3)
COMMON/OFFSET/C(3)
COMMON/XPROP/XPNEW(6)
DATA LF,LT/F,T/
DIMENSION XERR(6)
99601 FORMAT(1H0,I4,'IMPROPER INDEX *GOFX')
      IF(L$L(12))TIM1=TS
      TAU=TF-TIM1
      TAU=TF-TS
      TRAT1=(6.000/TAU**2)*(1.000-2.000*(TAU/TAU0))
      TRAT2=(2.000/TAU0)*(1.000-3.000*(TAU/TAU0))
      GOFX$=0.000
      RETURN
      ENTRY GOFX(J)
      GO TO (99999,99999,99999,99998,99998,99998),J
      WRITE(6,99601)J
      L$E(3)=LT
      GOFX=0.000
      RETURN
99999 GOFX=0.000
      RETURN
99998 F(J-3)=TRAT1*XP(J-3)+TRAT2*XP(J)
      GOFX=F(J-3)
      RETURN
      ENTRY GXINIT(J)
      WRITE(6,500)(XT(I),I=1,6)
500  FORMAT(1H1,3X,'XTRUE INITIALLY'/4X,1P6D12.5/)
      XERR(1)=2.00-02
      XERR(2)=1.50-02
      XERR(3)=1.00-02
      XERR(4)=7.00-03
      XERR(5)=5.00-03
      XERR(6)=4.00-04
      WRITE(6,501)(XERR(I),I=1,6)
501  FORMAT(1H0,3X,'INITIAL ERROR ON X'/4X,1P6D12.5/)
      DO 90001 I=1,ISLEN
90001 XP(I)=XT(I)+XERR(I)
      WRITE(6,776)(XP(I),I=1,ISLEN)
776  FORMAT(1H0,3X,'XHAT INITIALLY'/4X,1P6D12.5/)
      DO 25 I=1,6
25  XPNEW(I)=XP(I)
      GXINIT=0.000
      RETURN
      END
      SUBROUTINE FILTER

```

```

IMPLICIT REAL*8(A-H,O-Z)
REAL G,BQ,PSI
LOGICAL*1 L$L,L$E,L$M,L$F,L$T
COMMON/V$RBLE/XT(6),XP(6),XE(6),Z(6),ZP(6),ZE(6)
COMMON/T$MER/DELT,DT,TIME,TI,TF,N,ISLEN,IOLN
COMMON/SY$TEM/L$L(40),L$E(10)
COMMON/M$NITR/L$M(20)
COMMON/NOI$E/IRAN(10),DG(10),RFRAF(6)
COMMON/KALMAN/RM(6,6),FILT(6,4),U(6),RN(4,4),B(4,6)
COMMON/OFFSET/C(3)
COMMON/MOOCH/T6(4,4)
COMMON/XPROP/XPNEW(6)
COMMON/YTRUE/YTRUE(6)
DIMENSION T1(6),YL(6),Y(6),T2(6,6),T3(6,6),T4(6,6),QP(
*6,6),
$T5(6,6),PHI(6,6)
DIMENSION DY(6),QEXT(6,6),RU(1),DDY(6)
DIMENSION BQ(4,6),PSI(6,6)
DIMENSION G(6,1)
DIMENSION LL(6),MM(6)
DATA LF,LT/F,T/
TAUK1=TF-TIME
TAUK=TAUK1+DELT
RAT=TAUK1/TAUK
RAT2=RAT*RAT
RAT3=RAT2*RAT
A=3.000*RAT2-2.000*RAT3
DIF=RAT-RAT2
F=TAUK1*DIF
D=3.000*RAT2-2.000*RAT
E=-6.000*DIF/TAUK
DO 31 I=1,3
PHI(I,I)=A
PHI(I,I+3)=F
PHI(I+3,I)=E
PHI(I+3,I+3)=D
31 CONTINUE
C   PREDICT X STATE
    IF(L$L(9)) GO TO 8000
    DO 10 I=1,ISLEN
        T1(I)=0.000
    DO 10 J=1,ISLEN
10   T1(I)=T1(I)+PHI(I,J)*XP(J)
    GO TO 8002
8000 CONTINUE
    CALL INTGR
    DO 8001 I=1,ISLEN
8001 T1(I)=XP(I)
8002 CONTINUE
C   TRANSFORM TO Y SYSTEM
    P1=T1(1)+C(1)

```

```
P2=T1(2)+C(2)
P3=T1(3)+C(3)
Q=DSQRT(P1**2+P2**2)
R=DSQRT(P1**2+P2**2+P3**2)
AU=P1*T1(4)+P2*T1(5)
V=AU+P3*T1(6)
W=((P1)*(T1(5)))-((P2)*(T1(4)))
Y(1)=R
Y(2)=DATAN2(P2,P1)
Y(3)=DATAN2(P3,Q)
Y(4)=V/R
Y(5)=W/(Q**2)
Y(6)=(Q**2*T1(6)-AU*P3)/(Q*R**2)
CL=DCOS(Y(2))
CB=DCOS(Y(3))
SL=DSIN(Y(2))
SB=DSIN(Y(3))
C  COMPUTE LEFT PARTIAL DERIVATIVE
T2(1,1)=CL*CB
T2(2,1)=SL*CB
T2(3,1)=SB
T2(4,1)=-Y(6)*CL*SB-Y(5)*SL*CB
T2(5,1)=Y(5)*CL*CB-Y(6)*SL*SB
T2(6,1)=Y(6)*CB
T2(1,2)=-Y(1)*SL*CB
T2(2,2)=Y(1)*CL*CB
T2(4,2)=-Y(4)*SL*CB+Y(1)*Y(6)*SB*SL-Y(1)*Y(5)*CL*CB
T2(5,2)=Y(4)*CL*CB-Y(1)*Y(5)*SL*CB-Y(1)*Y(6)*CL*SB
T2(1,3)=-Y(1)*CL*SB
T2(2,3)=-Y(1)*SL*SB
T2(3,3)=Y(1)*CB
T2(4,3)=-Y(4)*CL*SB-Y(1)*Y(6)*CL*CB+Y(1)*Y(5)*SL*SB
T2(5,3)=-Y(4)*SL*SB-Y(1)*Y(5)*CL*SB-Y(1)*Y(6)*SL*CB
T2(6,3)=Y(4)*CB-Y(1)*Y(6)*SB
T2(4,4)=CB*CL
T2(5,4)=SL*CB
T2(6,4)=SB
T2(4,5)=-Y(1)*SL*CB
T2(5,5)=Y(1)*CL*CB
T2(4,6)=-Y(1)*CL*SB
T2(5,6)=-Y(1)*SL*SB
T2(6,6)=Y(1)*CB
C  COMPUTE RIGHT PARTIAL DERIVATIVE
T3(1,1)=P1/R
T3(1,2)=P2/R
T3(1,3)=P3/R
T3(2,1)=-P2/(Q**2)
T3(2,2)=P1/(Q**2)
T3(3,1)=-P1*P3/(Q*R**2)
T3(3,2)=-P2*P3/(Q*R**2)
T3(3,3)=Q/(R**2)
```

```

T3(4,1)=(T1(4)*R**2-V*P1)/(R**3)
T3(4,2)=(R**2*T1(5)-V*P2)/(R**3)
T3(4,3)=(T1(6)*R**2-V*P3)/(R**3)
T3(4,4)=P1/R
T3(4,5)=P2/R
T3(4,6)=P3/R
T3(5,1)=(T1(5)/(Q**2))-(2.*W*P1)/(Q**4)
T3(5,2)=(-T1(4)/(Q**2))-(2.*W*P2)/(Q**4)
T3(5,4)=-P2/(Q**2)
T3(5,5)=P1/(Q**2)
T3(6,1)=(P1*T1(6)-P3*T1(4))/(Q*R**2)-2.*Q*P1*T1(6)/(R*
**4)+
$2.*AU*P1*P3/(Q*R**4)+P1*P3*AU/(Q**3*R**2)
T3(6,2)=(P2*T1(6)-P3*T1(5))/(Q*R**2)-2.*Q*P2*T1(6)/(R*
**4)+
$2.*AU*P2*P3/(Q*R**4)+AU*P2*P3/(Q**3*R**2)
T3(6,3)=-AU/(Q*R**2)-2.*P3*T1(6)*Q/(R**4)+2.*AU*P3**2/
$(Q*R**4)
T3(6,4)=-P1*P3/(Q*R**2)
T3(6,5)=-P2*P3/(Q*R**2)
T3(6,6)=Q/(R**2)

```

```

C   COMPUTE PSI
      DO 12 I=1, ISLEN
      DO 12 J=1, ISLEN
      T4(I,J)=0.000
      DO 12 K=1, ISLEN
      DO 12 L=1, ISLEN
12   T4(I,J)=T4(I,J)+T2(I,K)*PHI(K,L)*T3(L,J)
      DO 9990 I=1,6
      DO 9990 J=1,6
      G(I,1)=0.0
9990 PSI(I,J)=T4(I,J)
      DO 9991 I=1,4
      DO 9991 J=1,6
9991 BQ(I,J)=B(I,J)

```

```

C   PREDICT COVARIANCE
      DO 13 I=1, ISLEN
      DO 13 J=1, ISLEN
      T5(I,J)=QP(I,J)
      DO 13 K=1, ISLEN
      DO 13 L=1, ISLEN
13   T5(I,J)=T5(I,J)+T4(I,K)*RM(K,L)*T4(J,L)
      DY(1)=-2.00-05
      DY(2)=-4.10-05
      DY(3)=-3.30-05
      DY(4)=1.50-05
      DY(5)=9.210-05
      DY(6)=4.10-05
      DO 600 I=1,6
      DO 600 J=1,6
600  QEXT(I,J)=DY(I)*DY(J)

```

```
      DO 980 I=1,6
980  QEXT(I,I)=QEXT(I,I)*1.0D6
      DO 601 I=1,6
      DO 601 J=1,6
601  T5(I,J)=T5(I,J)+QEXT(I,J)
C    COMPUTE THE FILTER
      DO 14 I=1,IOLEN
      DO 14 J=1,IOLEN
      T6(I,J)=RN(I,J)
      DO 14 K=1,ISLEN
      DO 14 L=1,ISLEN
14   T6(I,J)=T6(I,J)+B(I,K)*T5(K,L)*B(J,L)
      CALL MINV(T6,IOLEN,16,LL,MM,D)
      DO 15 I=1,ISLEN
      DO 15 J=1,IOLEN
      FILT(I,J)=0.0D0
      DO 15 K=1,ISLEN
      DO 15 L=1,IOLEN
15   FILT(I,J)=FILT(I,J)+T5(I,K)*B(L,K)*T6(L,J)
C    COMPUTE PREDICTED OBSERVATIONS
      DO 16 I=1,IOLEN
      ZP(I)=0.0D0
      DO 16 J=1,ISLEN
16   ZP(I)=ZP(I)+B(I,J)*Y(J)
C    COMPUTE THE ERROR IN OBSERVATIONS
      DO 17 I=1,IOLEN
17   ZE(I)=Z(I)-ZP(I)
C    UPDATE Y
      DO 18 I=1,ISLEN
      T1(I)=Y(I)
      DO 18 J=1,IOLEN
18   T1(I)=T1(I)+FILT(I,J)*ZE(J)
C    UPDATE COVARIANCE
      DO 19 I=1,ISLEN
      DO 19 J=1,ISLEN
      T4(I,J)=0.0D0
      DO 19 K=1,IOLEN
19   T4(I,J)=T4(I,J)+FILT(I,K)*B(K,J)
      DO 20 I=1,ISLEN
      DO 20 J=1,ISLEN
      T4(I,J)=-T4(I,J)
      IF(I.EQ.J) T4(I,J)=T4(I,J)+1.0D0
20  CONTINUE
      DO 21 I=1,ISLEN
      DO 21 J=1,ISLEN
      RM(I,J)=0.0D0
      DO 21 K=1,ISLEN
21   RM(I,J)=RM(I,J)+T4(I,K)*T5(K,J)
99  CONTINUE
C    SAVE Y(K+1,K+1)
      DO 22 I=1,ISLEN
```

```

22 YL(1)=T1(I)
   CALL CONOBS(6,1,4,PSI,G,BQ)
C   CONVERT TO X
      CL=DCOS(YL(2))
      CB=DCOS(YL(3))
      SL=DSIN(YL(2))
      SB=DSIN(YL(3))
      XP(1)=YL(1)*CB*CL-C(1)
      XP(2)=YL(1)*CB*SL-C(2)
      XP(3)= YL(1)*SB-C(3)
      XP(4)=YL(4)*CB*CL-YL(1)*YL(6)*SB*CL-YL(1)*YL(5)*CB*SL
      XP(5)=YL(4)*CB*SL-YL(1)*YL(6)*SB*SL+YL(1)*YL(5)*CB*CL
      XP(6)=YL(4)*SB+YL(1)*YL(6)*CB
      WRITE(6,912)(XP(J),J=1,6)
912 FORMAT(1H0,3X,'XHAT'/4X,1P6D12.5/)
      DO 913 I=1,6
913  XPNEW(I)=XP(I)
      RETURN
      ENTRY FLINIT
      *          ENTRY
C   ZERO ARRAYS
      DO 101 I=1,ISLEN
      DO 102 J=1,ISLEN
        T2(I,J)=0.0D0
        T3(I,J)=0.0D0
        RM(I,J)=0.0D0
        PHI(I,J)=0.0D0
102  QP(I,J)=0.0D0
      DO 101 J=1,IOLEN
101  FILT(I,J)=0.0D0
C   INITIALIZE VARIABLES
      YL(1)=DSQRT((XP(1)+C(1))**2+(XP(2)+C(2))**2+(XP(3)+C(3)
        *)**2)
      YL(2)=DATAN2((XP(2)+C(2)),(XP(1)+C(1)))
      ARG2=DSQRT((XP(1)+C(1))**2+(XP(2)+C(2))**2)
      YL(3)=DATAN2((XP(3)+C(3)),ARG2)
      YL(4)=((XP(1)+C(1))*XP(4)+(XP(2)+C(2))*XP(5)+(XP(3)+C(
        *3))*XP(6))
      $/YL(1)
      YL(5)=((XP(1)+C(1))*XP(5)-(XP(2)+C(2))*XP(4))/
      $((XP(1)+C(1))**2+(XP(2)+C(2))**2)
      YL(6)=(((XP(1)+C(1))**2+(XP(2)+C(2))**2)*XP(6)-(XP(3)+
        *C(3))*
      $((XP(1)+C(1))*XP(4)+(XP(2)+C(2))*XP(5)))/(YL(1)**2*ARG
        *2)
      RM(1,1)=1.0D0
      RM(2,2)=1.0D0
      RM(3,3)=1.0D0
      RM(4,4)=1.0D0
      RM(5,5)=1.0D0
      RM(6,6)=1.0D0

```

```

WRITE(6,652)
652 FORMAT(1H0,3X,'INITIAL COVARIANCE MATRIX')
DO 650 I=1,6
650 WRITE(6,651)(RM(I,J),J=1,6)
651 FORMAT(1H0,4X,1P6D12.5)
RETURN
END
SUBROUTINE OBSERV
IMPLICIT REAL*8(A-H,O-Z)
LOGICAL*1 L$L,L$E,L$MON,LF,LT
COMMON/V$RBLE/XT(6),XP(6),XE(6),Z(6),ZP(6),ZE(6)
COMMON/T$MER/DELT,DT,TIME,TI,TF,N,ISLEN,IOLEN
COMMON/SY$TEM/L$L(40),L$E(10)
COMMON/M$NITR/L$MON(20)
COMMON/NOI$E/IRAN(10),DG(10),RFRAF(6)
COMMON/KALMAN/CRUD(66),RN(4,4),B(4,6)
COMMON/OFFSET/C(3)
COMMON/YTRUE/YTRUE(6)
DIMENSION ZT(6)
DATA LF,LT/F,T/
C OBSERVATIONS
C RFRAF(1 TO 4) = STDDEV FOR ZT(1 TO 4) RESPECTIVELY
YTRUE(1)=DSQRT((XT(1)+C(1))**2+(XT(2)+C(2))**2+(XT(3)+
*C(3))**2)
RANGE=YTRUE(1)
YTRUE(2)=DATAN2((XT(2)+C(2)),(XT(1)+C(1)))
ARG1=DSQRT((XT(1)+C(1))**2+(XT(2)+C(2))**2)
YTRUE(3)=DATAN2((XT(3)+C(3)),ARG1)
YTRUE(4)=((XT(1)+C(1))*XT(4)+(XT(2)+C(2))*XT(5)+(XT(3)
*+C(3))*XT(6)
$)/YTRUE(1)
YTRUE(5)=((XT(1)+C(1))*XT(5)-(XT(2)+C(2))*XT(4))/(ARG1
**2)
YTRUE(6)=((ARG1**2)*XT(6)-(XT(3)+C(3))*((XT(1)+C(1))*X
*T(4)
$+(XT(2)+C(2))*XT(5)))/(YTRUE(1)**2*ARG1)
RFRAF(1)=DABS(YTRUE(1)*3.0D-03)
RFRAF(2)=EANG
RFRAF(3)=EANG
RFRAF(4)=1.0D-02
DG(1)=GRAND(0.0D0,RFRAF(1),5)
DG(2)=GRAND(0.0D0,RFRAF(2),2)
DG(3)=GRAND(0.0D0,RFRAF(3),3)
DG(4)=GRAND(0.0D0,RFRAF(4),8)
IF(RANGE.GT.RMIN)GO TO 2
IF(RANGE.LT.RMIN.AND.RANGE.GT.RDMIN)GO TO 3
IF(RANGE.LT.RDMIN.AND.RANGE.GT.RMIN) GO TO 6
IF(RANGE.LT.RDMIN)GO TO 8
2 WRITE(6,1)(DG(LSCVT),LSCVT=2,3)
IOLEN=2
GO TO 4

```



```
3 IOLEN=3
  WRITE(6,1)(DG(LSCNT),LSCNT=1,IOLEN)
1 FORMAT(///,10X,'NOISE',5X,1P4D12.5)
  GO TO 4
6 WRITE(6,1)(DG(LSCNT),LSCNT=2,4)
  IOLEN=3
  GO TO 4
8 IOLEN=4
  WRITE(6,1)(DG(LSCNT),LSCNT=1,IOLEN)
4 CONTINUE
  DO 38 I=1,IOLEN
  DO 32 J=1,IOLEN
32 RN(I,J)=0.0D0
  DO 38 J=1,IOLEN
38 B(I,J)=0.0D0
  IF(IOLEN.EQ.2)GO TO 34
  IF(L$L(20).AND.IOLEN.EQ.3)GO TO 37
  DO 33 I=1,IOLEN
33 B(I,1)=1.0D0
  GO TO 36
37 B(1,2)=1.0D0
  B(2,3)=1.0D0
  B(3,4)=1.0D0
  DO 39 I=1,3
39 RN(I,1)=RFRAF(I+1)**2
  GO TO 35
34 B(1,2)=1.0D0
  B(2,3)=1.0D0
  RN(1,1)=RFRAF(2)**2
  RN(2,2)=RFRAF(3)**2
  GO TO 35
36 DO 30 I=1,IOLEN
30 RN(I,1)=RFRAF(I)**2
35 CONTINUE
  DO 41 I=1,IOLEN
  ZT(I)=0.0D0
  DO 41 J=1,IOLEN
41 ZT(I)=ZT(I)+B(I,J)*YTRUE(J)
  DO 43 I=1,IOLEN
43 Z(I)=ZT(I)+DG(I)
  RETURN
  ENTRY OVINIT
*          ENTRY
  A=0.0D0
  EANG=2.906D-04
  RMIN=1.0D20
  RDMIN=1.0D20
  RETURN
  END
  FUNCTION GRAND(RMEAN,STDDEV,ISLCT)
  IMPLICIT REAL*8(A-H,O-Z)
```

```

C  PURPOSE
C  COMPUTES A NORMALLY DISTRIBUTED RANDOM NUMBER WITH A G
  *IVEN
C  MEAN AND STANDARD DEVIATION
  A=0.000
  DO 50 I=1,12
50  A=A+URAND(1SLCT)
  GRAND=(A-6.000)*STDDEV+RMEAN
  RETURN
  END
  FUNCTION URAND(1SLCT)
  IMPLICIT REAL*8(A-H,O-Z)
  COMMON/NOISE/IRAN(10),DG(10),RFRAF(6)
  IY=IRAN(1SLCT)*65539
  IF(IY)5,6,6
5  IY=IY+2147483647+1
6  URAND=DFLOAT(IY)*4.656613D-10
  IRAN(1SLCT)=IY
  RETURN
  ENTRY URINIT(1SLCT)
  *      ENTRY
  URAND=0.000
  URINIT=0.000
  RETURN
  END
  BLOCK DATA
  IMPLICIT REAL*8 (A-H,O-Z)
  COMMON/NOISE/IRAN(10),DG(10),RFRAF(6)
  DATA IRAN/69800661,54218059,51070625,15239339,75892237
  *,
  *10418327,81767867,59847821,52031357,26256073/
  END
  SUBROUTINE CYCLOT
  IMPLICIT REAL*8(A-H,O-Z)
  REAL GUTL
  LOGICAL*1 L$L,L$E,LMON,LF,LT
  COMMON/V$RBLE/XT(6),XP(6),XE(6),Z(6),ZP(6),ZE(6)
  COMMON/T$MER/DELT,DT,TIME,TI,TF,N,ISLEN,IOLN
  COMMON/SYSTEM/L$L(40),L$E(10)
  COMMON/M$NITR/LMON(20)
  COMMON/NOISE/IRAN(10),DG(10),RFRAF(6)
  COMMON/KALMAN/P(6,6),FILT(6,4),U(6),RN(4,4),B(4,6)
  COMMON/FORCE/F(3)
  COMMON/OFFSET/C(3)
  COMMON/MOOCH/T6(4,4)
  DATA LF,LT/F,T/
601 FORMAT(1H ,3X,'TRUE STATE VECTOR'/4X,1PD12.5)
603 FORMAT(1H , 'TIME=',1PD12.5)
604 FORMAT(1H , 'NORMED DISTANCE=',1PD12.5,3X,'NORMED VELOC
  *ITY=',
  11PD12.5,3X,'NORMED FORCE=',1PD12.5)

```

```

605 FORMAT(1H0,'$$$ RENDEZVOUS $$$')
606 FORMAT(1H0,'MINIMUM NORMED DISTANCE=',1PD12.5,3X,'AT T
    *IME=',
    1 1PD12.5)
607 FORMAT(1H0,'MINIMUM NORMED VELOCITY =',1PD12.5,3X,'AT
    *TIME=',
    1 1PD12.5)
608 FORMAT(////)
609 FORMAT(1H0,'**** OUT OF TIME ****')
610 FORMAT(1H0,'FORCE VECTOR',/,1H ,1P3D12.5)
611 FORMAT(1H0,3X,'PREDICTED STATE VECTOR'/4X,1P6D12.5)
612 FORMAT(1H0,3X,'ERROR IN STATE VECTOR'/4X,1P6D12.5)
613 FORMAT(1H0,3X,'TRUE OBSERVATIONS'/4X,1P6D12.5)
614 FORMAT(1H0,3X,'PREDICTED OBSERVATIONS'/4X,1P6D12.5)
615 FORMAT(1H0,3X,'RESIDUAL ERROR'/4X,1P6D12.5)
617 FORMAT(1H0,3X,'COVARIANCE MATRIX')
618 FORMAT(1H ,6X,1P6D12.5)
619 FORMAT(1H1,10X,' SIMULATION RESULTS',////)
    IF(L$L(1))WRITE(6,619)
620 FORMAT(1H ,3X,'NORMED POSITION ERROR = ',1PD12.5,5X,'
    *NORMED VELOC
    *ITY ERROR = ',1PD12.5)
    FT=0.0D0
    T1=0.0D0
    T2=0.0D0
    DO 1 I=1,3
    F(I)=F(I)*CF1
    FT=FT+F(I)**2
    T1=T1+XT(I)*XT(I)
1  T2=T2+XT(I+3)*XT(I+3)
    FT=DSQRT(FT)
    XS=DSQRT(T1)
    XV=DSQRT(T2)
    IF(L$L(9))GO TO 2
    IF(XS.GT.XO)GO TO 3
    XO=XS
    TXS=TIME
3  IF(XV.GT.XVO)GO TO 2
    XVO=XV
    TXV=TIME
2  WRITE(6,603)TIME
    WRITE(6,604)XS,XV,FT
    CHK=DABS(TF-TIME)
    IF(CHK.LE.(1.10D0*DELT)) GO TO 50
    GO TO 60
50  FTHRST=DSQRT(F(1)**2+F(2)**2+F(3)**2)
    GUTL=FTHRST/1.0D-04
    GUTTR=DFLOAT(IFIX(GUTL))
    QCHK=GUTTR+1.0D0
    TF=TF+QCHK*DELT
60  CONTINUE

```

```

IF(L$L(2))RETURN
DO 99901 I$=1,ISLEN
99901 XE(I$)=XT(I$)-XP(I$)
PNORM=DSQRT(XE(1)**2+XE(2)**2+XE(3)**2)
VNORM=DSQRT(XE(4)**2+XE(5)**2+XE(6)**2)
WRITE(6,610)F
WRITE(6,601)XT
WRITE(6,611)XP
WRITE(6,612)XE
WRITE(6,620)PNORM,VNORM
WRITE(6,613)Z
WRITE(6,614)ZP
WRITE(6,615)ZE
WRITE(6,617)
DO 6 I=1,ISLEN
6 WRITE(6,618)(P(I,J),J=1,ISLEN)
WRITE(6,608)
99902 L$L(1)=LF
RETURN
ENTRY TERMIN
* ENTRY
IF(XS.LT.1.0D-2.AND.XV.LT.1.0D-4)GO TO 4
IF(TIME.GE.TF)GO TO 5
RETURN
4 WRITE(6,605)
L$L(6)=LF
RETURN
5 WRITE(6,609)
L$L(6)=LF
RETURN
ENTRY RECAP
* ENTRY
WRITE(6,606)XO,TXS
WRITE(6,607)XVO,TVX
RETURN
ENTRY CYINIT
XO=1.0D40
XVO=1.0D40
C GRAVITATIONAL ACCELERATION INVERSE
CF1=1.0D0/(9.80665D-08*8.64D4**2)
RETURN
END
SUBROUTINE TARGET(S,T)
IMPLICIT REAL*8 (A-H,O-Z)
DIMENSION S(3)
COMMON/VAR1/A,RN,EPS,ET,TSTAR,DET
601 FORMAT(1H0,'CONVERGENCE= ',1PD12.5,' KEPEQ')
EO=ET
ET=ET+DET
DO 1 I=1,100
SINET=DSIN(ET)

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```

COSET=DCOS(ET)
F=RN*(T+TSTAR)-ET+EPS*SINET
DF=EPS*COSET-1.000
ET=ET-F/DF
DIF=DABS(F/DF)
IF(DIF.LT.1.00-10)GO TO 2
1 CONTINUE
WRITE(6,601)DIF
2 DET=ET-EO
S(1)=A*(COSET -EPS)
S(2)=A*DSQRT(1.000-EPS*EPS)*SINET
S(3)=0.000
RETURN
END
SUBROUTINE MINV(A,N,NSQ,L,M,BIGA)
IMPLICIT REAL*8 (A-H,O-Z)
*          00000135
  DIMENSION A(NSQ),L(6),M(6)

C          *          00013200
C          *          DESCRIPTION OF PARAMETERS
C          *          00013500
C          *          A - INPUT MATRIX, DESTROYED IN COMPUTATION AND R
C          *          *EPLACED BY 00013600
C          *          RESULTANT INVERSE.
C          *          00013700
C          *          N - ORDER OF MATRIX A
C          *          00013800
C          *          BIGA - RESULTANT DETERMINANT
C          *          00013900
C          *          L - WORK VECTOR OF LENGTH N
C          *          00014000
C          *          M - WORK VECTOR OF LENGTH N
C          *          00014100
C          *          00014200
C          *          NK=-N
C          *          00017500
C          *          DO 190 K=1,N
C          *          00017600
C          *          NK=NK+N
C          *          00017700
C          *          L(K)=K
C          *          00017800
C          *          M(K)=K
C          *          00017900
C          *          KK=NK+K
C          *          00018000
C          *          BIGA=A(KK)
C          *          00018100
C          *          DO 30 J=K,N
C          *          00018200
C          *          IZ=N*(J-1)

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```

*          00018300
DO 30 I=K,N
*          00018400
IJ=IZ+I
*          00018500
10 IF(DABS(BIGA)-DABS(A(IJ))) 20,30,30
20 BIGA=A(IJ)
*          00018800
L(K)=I
*          00018900
M(K)=J
*          00019000
30 CONTINUE
*          00019100
C
*          00019200
C      INTERCHANGE ROWS
*          00019300
C
*          00019400
J=L(K)
*          00019500
IF(J-K) 60,60,40
*          00019600
40 KI=K-N
*          00019700
DO 50 I=1,N
*          00019800
KI=KI+N
*          00019900
HOLD=-A(KI)
*          00020000
JI=KI-K+J
*          00020100
A(KI)=A(JI)
*          00020200
50 A(JI)=HOLD
*          00020300
C
*          00020400
C      INTERCHANGE COLUMNS
*          00020500
C
*          00020600
60 I=M(K)
*          00020700
IF(I-K) 90,90,70
*          00020800
70 JP=N*(I-1)
*          00020900
DO 80 J=1,N
*          00021000
JK=NK+J
*          00021100
JI=JP+J
*          00021200
HOLD=-A(JK)
*          00021300
A(JK)=A(JI)
*          00021400

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```

80 A(JI) =HOLD
*      00021500
C
*      00021600
C      DIVIDE COLUMN BY MINUS PIVOT (VALUE OF PIVOT ELEMEN
*T IS 00021700
C      CONTAINED IN BIGA)
*      00021800
C
*      00021900
90 IF(BIGA) 110,100,110
*      00022000
100 RETURN
*      00022100
110 DO 130 I=1,N
*      00022200
    IF(I-K) 120,130,120
*      00022300
120 IK=NK+I
*      00022400
    A(IK)=A(IK)/(-BIGA)
*      00022500
130 CONTINUE
*      00022600
C
*      00022700
C      REDUCE MATRIX
*      00022800
C
*      00022900
    DO 160 I=1,N
*      00023000
        IK=NK+I
*      00023100
        IJ=I-N
*      00023200
        DO 160 J=1,N
*      00023300
            IJ=IJ+N
*      00023400
            IF(I-K) 140,160,140
*      00023500
140 IF(J-K) 150,160,150
*      00023600
150 KJ=IJ-I+K
*      00023700
    A(IJ)=A(IK)*A(KJ)+A(IJ)
*      00023800
160 CONTINUE
*      00023900
C
*      00024000
C      DIVIDE ROW BY PIVOT
*      00024100
C
*      00024200
    KJ=K-N
*      00024300
    DO 180 J=1,N
*      00024400

```

```
      KJ=KJ+N
      *          00024500
      IF(J-K) 170,180,170
      *          00024600
170  A(KJ)=A(KJ)/BIGA
      *          00024700
180  CONTINUE
      *          00024800
C
      *          00024900
C      REPLACE PIVOT BY RECIPROCAL
      *          00025000
C
      *          00025100
      A(KK)=1.0/BIGA
      *          00025200
190  CONTINUE
      *          00025300
C
      *          00025400
C      FINAL ROW AND COLUMN INTERCHANGE
      *          00025500
C
      *          00025600
      K=N
      *          00025700
200  K=(K-1)
      *          00025800
      IF(K) 270,270,210
      *          00025900
210  I=L(K)
      *          00026000
      IF(I-K) 240,240,220
      *          00026100
220  JQ=N*(K-1)
      *          00026200
      JR=N*(I-1)
      *          00026300
      DO 230 J=1,N
      *          00026400
      JK=JQ+J
      *          00026500
      HOLD=A(JK)
      *          00026600
      JI=JR+J
      *          00026700
      A(JK)=-A(JI)
      *          00026800
230  A(JI) =HOLD
      *          00026900
240  J=M(K)
      *          00027000
      IF(J-K) 200,200,250
      *          00027100
250  KI=K-N
      *          00027200
      DO 260 I=1,N
      *          00027300
      KI=KI+N
      *          00027400
```



```
HOLD=A(KI)
*          00027500
JI=KI-K+J
*          00027600
A(KI)=-A(JI)
*          00027700
260 A(JI) =HOLD
*          00027800
GO TO 200
*          00027900
270 RETURN
*          00028000
END
*          00028100
SUBROUTINE CONOBS (NP,NU,NY,F,G,H)
*          CNBS0000
DIMENSION F(6,6),G(6,1),H(4,6),B(6,24),S1(6,6),W(6,6),
*S(6,6)
DOUBLE PRECISION DET1
LO=NU
*          CNBS0050
DO 32 I=1,NP
*          CNBS0060
DO 32 J=1,NU
*          CNBS0070
32  S(I,J) = G(I,J)
*          CNBS0080
DO 33 I=1,NP
*          CNBS0090
DO 33 J=1,NP
*          CNBS0100
33  S1(I,J)=F(I,J)
*          CNBS0110
DO 85 ITEST=1,2
*          CNBS0120
IF (LO .EQ. 0) GO TO 90
*          CNBS0130
CALL NORML (NP,LO,S)
*          CNBS0140
DO 35 I = 1,NP
*          CNBS0150
DO 35 J = 1,LO
*          CNBS0160
35  B(I,J) = S(I,J)
*          CNBS0170
L=1
*          CNBS0180
MOU=NP-1
*          CNBS0190
DO 40 IT=1,MOU
*          CNBS0200
CALL MMUL (NP,NP,LO,S1,S,W)
*          CNBS0210
CALL NORML (NP,LO,W)
*          CNBS0220
DO 20 I=1,NP
*          CNBS0230
DO 20 J=1,LO
*          CNBS0240
JI=(J+L*LO)
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*          CNBS0250
  B(I,J1)=W(I,J)
*          CNBS0260
20 S(I,J)=W(I,J)
*          CNBS0270
40 L=L+1
*          CNBS0280
  NDIM=L*LO
*          CNBS0290
  DO 41 I = 1,NP
*          CNBS0292
    SUM = 0.0
*          CNBS0294
    DO 42 J = 1,NDIM
*          CNBS0296
42 SUM = SUM + B(I,J)**2
*          CNBS0298
    SUM = SQRT(SUM)
*          CNBS0300
    IF (SUM .EQ. 0.) GO TO 41
*          CNBS0302
    DO 45 J = 1,NDIM
*          CNBS0304
45 B(I,J) = B(I,J)/SUM
*          CNBS0306
41 CONTINUE
*          CNBS0308
  DO 1000 II = 1,NP
*          CNBS0310
  DO 1000 JJ = 1,NP
*          CNBS0320
    SUM = 0.0
*          CNBS0330
    DO 1001 KK = 1,NDIM
*          CNBS0340
1001 SUM = SUM + B(II,KK)*B(JJ,KK)
*          CNBS0350
1000 W(II,JJ) = SUM
*          CNBS0360
  CALL MXINV (NP,W,DET1,IER)
  IF (ITEST .GT. 1) GO TO 87
*          CNBS0380
  IF (IER .EQ. 0) GO TO 70
*          CNBS0390
  GO TO 90
*          CNBS0420
70 WRITE(6,300) DET1
300 FORMAT (26H CONTROLLABLE          DET = ,1PE14.4)
*          CNBS0440
90 LO=NY
*          CNBS0450
  DO 93 J=1,NY
*          CNBS0460
  DO 93 I=1,NP
*          CNBS0470
93 S(I,J)=H(J,I)
*          CNBS0480
85 CALL MTRANS (NP,NP,F,S1)
*          CNBS0490
87 IF (IER .EQ. 0) GO TO 170

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*          CNBS0500
WRITE(6,902) DET1
902 FORMAT (26H NOT OBSERVABLE      DET = ,1PE14.4)
*          CNBS0520
GO TO 91
*          CNBS0530
170 WRITE(6,901) DET1
901 FORMAT (26H OBSERVABLE          DET = ,1PE14.4)
*          CNBS0550
91 RETURN
*          CNBS0560
END
*          CNBS0570
SUBROUTINE MATRIX
*          MTRX0000
DIMENSION A(6,6),B(6,6),C(6,6)
ENTRY      MMUL (MP,NP,NU,A,B,C)
*          MTRX0520
DO 11 L=1,MP
*          MTRX0530
DO 11 I=1,NU
*          MTRX0540
SUM = 0.0
*          MTRX0550
DO 31 J=1,NP
*          MTRX0560
31 SUM = SUM + A(L,J)*B(J,I)
*          MTRX0570
11 C(L,I) = SUM
*          MTRX0580
RETURN
*          MTRX0590
ENTRY      MTRANS (M,N,A,B)
*          MTRX0790
DO 10 I=1,M
*          MTRX0800
DO 10 J=1,N
*          MTRX0810
10 B(J,I) = A(I,J)
*          MTRX0820
RETURN
*          MTRX0830
ENTRY      MXINV (NP,A,DET1,IER)
DOUBLE PRECISION DA(6,6),DET1,DB(1,1)
DO 1 I = 1,NP
*          MTRX0860
DO 1 J = 1,NP
*          MTRX0870
1 DA(I,J) = A(I,J)
*          MTRX0880
CALL DINVER (NP,6,DA,0,1,DB,DET1,IER)
IF (IER .NE. 0) RETURN
*          MTRX0900
DO 2 I = 1,NP
*          MTRX0910
DO 2 J = 1,NP
*          MTRX0920
2 A(I,J) = DA(I,J)
*          MTRX0930
RETURN

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*           MTRX0940
ENTRY NORML (NR,NC,A)
*           MTRX1000
DO 4 J = 1,NC
*           MTRX1010
SUM = 0.0
*           MTRX1020
DO 3 I = 1,NR
*           MTRX1030
SUM = SUM + A(I,J)**2
3 *           MTRX1040
SUM = SQRT(SUM)
*           MTRX1050
IF (SUM .EQ. 0.) GO TO 4
*           MTRX1060
DO 5 I = 1,NR
*           MTRX1070
5 A(I,J) = A(I,J)/SUM
*           MTRX1080
4 CONTINUE
*           MTRX1090
RETURN
*           MTRX1100
END
*           MTRX1110
SUBROUTINE DINVER (NA,NAD,A,NB,NBD,B,DET1,IERROR)
C THIS SUBROUTINE IS A MODIFICATION OF THE UNIVERSITY OF
* FLORIDA DNVRO010
C COMPUTER CENTER'S INVERT . IT USES DOUBLE PRECISION AN
* D HAS BEEN DNVRO020
C RENAMED DINVER. C FOSHA 2-69
* DNVRO030
DIMENSION A(NAD,NAD),B(NBD,NBD),BD(6),INDEX(6)
DOUBLE PRECISION A,B,BD,SAVE,PIVOT,DET1
DET1=1.000
IERROR = 0
* DNVRO070
DO 130 I = 1, NA
* DNVRO080
PIVOT = 0.000
* DNVRO090
C SEARCH FOR PIVOTAL ELEMENT
* DNVRO100
DO 60 J = I, NA
* DNVRO110
IF (DABS(A(J,I)) .LE. DABS(PIVOT)) GO TO 60
* DNVRO120
PIVOT = A(J,I)
* DNVRO130
INDEX(I) = J
* DNVRO140
60 CONTINUE
* DNVRO150
IF (DABS(PIVOT) .LT. 1. D-6) GO TO 250
* DNVRO160
IF (INDEX(I) .EQ. I) GO TO 90
* DNVRO170
DET1=-DET1
C INTERCHANGE ROWS TO PUT PIVOTAL ELEMENT ON D
* IAGONAL DNVRO190

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DO 80 L = 1, NA
*      DNVR0200
*      SAVE = A(I,L)
*      DNVR0210
*      A(I,L) = A(INDEX(I), L)
*      DNVR0220
80    A(INDEX(I), L) = SAVE
*      DNVR0230
90    DET1=DET1*PIVOT
*      A(I,I) = 1.000
*      DNVR0250
DO 91 KK=1,NA
*      DNVR0260
91    A(I,KK)=A(I,KK)/PIVOT
*      DNVR0270
C      REDUCE NON-PIVOTAL ROWS
*      DNVR0280
DO 130 LJ = 1, NA
*      DNVR0290
*      IF (LJ .EQ. I) GO TO 130
*      DNVR0300
*      SAVE = A(LJ, I)
*      DNVR0310
*      A(LJ,I) = 0.000
*      DNVR0320
DO 120 K = 1, NA
*      DNVR0330
120   A(LJ,K) = A(LJ,K) - SAVE * A(I,K)
*      DNVR0340
130   CONTINUE
*      DNVR0350
C      INTERCHANGE COLUMNS
*      DNVR0360
*      NA1=NA+1
*      DNVR0370
DO 160 KKK = 1, NA
*      DNVR0380
*      K = NA1 - KKK
*      DNVR0390
*      IF (INDEX(K) .EQ. K) GO TO 160
*      DNVR0400
DO 98 L = 1, NA
*      DNVR0410
*      SAVE = A(L,K)
*      DNVR0420
*      A(L,K) = A(L, INDEX(K))
*      DNVR0430
98    A(L, INDEX(K)) = SAVE
*      DNVR0440
160   CONTINUE
*      DNVR0450
C      A INVERSE IS NOW STORED IN A
*      DNVR0460
C      FIND SOLUTION VECTORS FOR ALL CONSTANT VECTO
*      DNVR0470
*      IF (NB .LE. 0) RETURN
*      DNVR0480
DO 190 K = 1, NB
*      DNVR0490
DO 180 I = 1, NA

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*          DNVR0500
  BD(I) = 0.0D0
*          DNVR0510
  DO 180 J = 1, NA
*          DNVR0520
180  BD(I) = BD(I) + A(I,J)*B(J,K)
*          DNVR0530
  DO 190 I = 1, NA
*          DNVR0540
190  B(I,K) = BD(I)
*          DNVR0550
C          SOLUTION VECTORS NOW IN B
*          DNVR0560
  RETURN
*          DNVR0570
C          IF CONTROL REACHES 250, MATRIX IS SINGULAR
*          DNVR0580
250  IERROR = +1
*          DNVR0590
  DET1=0.0D0
  RETURN
*          DNVR0610
  END
*          DNVR0620
  SUBROUTINE INTGR
    IMPLICIT REAL*8 (A-H,O-Z)
    LOGICAL*1 L$L,L$E,L$MON,LF,LT
    LOGICAL*1 LS1,LS2
    COMMON/V$RBLE/CRAP(6),XT(6),XE(6),Z(6),ZP(6),ZE(6)
    COMMON/T$MER/DELT,DT,TIME,TI,TF,N,ISLEN,IOLEN
    COMMON/SYSTEM/L$L(40),L$E(10)
    COMMON/M$NITR/L$MON(20)
    COMMON/NOI$E/IRAN(10),DG(10),RFRAF(6)
    COMMON/OFFSET/C(3)
    COMMON/XPROP/XPNEW(6)
    DIMENSION XINT(6),SUM(6)
    DATA LF,LT/F,T/
    L$L(12)=LT
700  CONTINUE
    TIME=TIME-DELT
    DO 1 ICYCLE=1,N
    DO 33 I=1,ISLEN
33  SUM(I)=0.0D0
    L$L(10)=LT
    DO 10 I1=1,4
    LS1=I1.EQ.2. OR.I1.EQ.3
    LS2=I1.EQ.4
    L$L(11)=I1.EQ.3
    F=F1
    FS=F5
    IF(LS1)F=F2
    IF(LS1)FS=F3
    IF(LS2)FS=F4
    TS=TIME+DT*FS
    DO 20 I=1,ISLEN
20  XINT(I)=XT(I)+FS*XINT(I)
    DO 31 I=1,NP1
    J=I-1
    IF(J.GT.0) GO TO 2
    DUM=DOFX$N(J,TS)

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DUM=GOFX$N(J,TS)
L$L(12)=LF
GO TO 31
2 XINT(J)=DT*(DOFXN(J)+GOFXN(J))
  SUM(J)=SUM(J)+F*XINT(J)
31 CONTINUE
  L$L(10)=LF
10 CONTINUE
  TIME=TIME+DT
  DO 11 I=1,ISLEN
11 XT(I)=XT(I)+SUM(I)
  1 CONTINUE
701 CONTINUE
  RETURN
  ENTRY INTINI
  F1=1.000/6.000
  F2=2.000*F1
  F3=1.000/2.000
  F4=1.000
  F5=0.000
  NP1=ISLEN+1
  DO 32 I=1,ISLEN
32 XINT(I)=XT(I)
  RETURN
  END
  FUNCTION DOFX$N(J,TS)
  IMPLICIT REAL*8(A-H,O-Z)
  LOGICAL*1 L$L,L$E,LMON,LF,LT
  COMMON/V$RBLE/CRAP(6),XT(6),XE(6),Z(6),ZP(6),ZE(6)
  COMMON/T$MER/DELT,DT,TIME,TI,TF,N,ISLEN,IOLEN
  COMMON/SYSTEM/L$L(40),L$E(10)
  COMMON/M$NITR/LMON(20)
  COMMON/NOI$E/IRAN(10),DG(10),RFRAF(6)
  COMMON/VAR1/A,RN,EPS,EO,TSTAR,TRASH
  COMMON/NEWDET/DETN
  COMMON/OFFSET/C(3)
  DIMENSION D(3),S(3)
  DATA LF,LT/F,T/
99601 FORMAT(1H0,I4,'IMPROPER INDEX *DOFX*')
  IF(.NOT.L$L(11))CALL TARGEN(S,TS)
  S2=0.000
  D2=0.000
  DO 1 I=1,3
  D(I)=S(I)+C(I)+XT(I)
  D2=D2+D(I)**2
  1 S2=S2+S(I)**2
  DN=DSQRT(D2)
  SN=DSQRT(S2)
  RAT1=GM/(SN*SN*SN)
  RAT2=(SN/DN)**3
  DOFX$N=0.000
  RETURN
  ENTRY DOFXN(J)
  GO TO (99999,99999,99999,99998,99998,99998),J
  WRITE(6,99601)J
  DOFXN=0.000
  L$E(2)=LT
  RETURN
99999 DOFXN=XT(J+3)
  RETURN

```

```

99998 DOFXN=RATI*(S(J-3)-D(J-3)*RAT2)
      RETURN
      ENTRY DXININ(J)
      GM=9.90549D05
      DETN=1.0D-3
      DXININ=0.000
      RETURN
      END
      FUNCTION GOFX$N(J,TS)
      IMPLICIT REAL*8(A-H,O-Z)
      LOGICAL*1 L$L,L$E,L$MON,LF,LT
      COMMON/V$RBLE/XT(6),XP(6),XE(6),Z(6),ZP(6),ZE(6)
      COMMON/T$MER/DELT,DT,TIME,TI,TF,N,ISLEN,I$LEN
      COMMON/SYSTEM/L$L(40),L$E(10)
      COMMON/M$NITR/L$MON(20)
      COMMON/NOISE/IRAN(10),DG(10),RFRAF(6)
      COMMON/FORCE/F(3)
      COMMON/OFFSET/C(3)
      COMMON/XPROP/XPNEW(6)
      DATA LF,LT/F,T/
99601 FORMAT(1H0,I4,'IMPROPER INDEX *GOFX')
      IF(L$L(12))TIM1=TS
      TAUO=TF-TIM1
      TAU=TF-TS
      TRAT1=(6.0D0/TAUO**2)*(1.0D0-2.0D0*(TAU/TAUO))
      TRAT2=(2.0D0/TAUO)*(1.0D0-3.0D0*(TAU/TAUO))
      GOFX$N=0.000
      RETURN
      ENTRY GOFXN(J)
      GO TO (99999,99999,99999,99998,99998,99998),J
      WRITE(6,99601)J
      L$E(3)=LT
      GOFXN=0.000
      RETURN
99999 GOFXN=0.000
      RETURN
99998 F(J-3)=TRAT1*XPNEW(J-3)+TRAT2*XPNEW(J)
      GOFXN=F(J-3)
      RETURN
      ENTRY GXININ(J)
      GXININ=0.000
      RETURN
      END
      SUBROUTINE TARGEN(S,T)
      IMPLICIT REAL*8 (A-H,O-Z)
      DIMENSION S(3)
      COMMON/VAR1/A,RN,EPS,ET,TSTAR,TRASH
      COMMON/NEWDET/DETN
601  FORMAT(1H0,'CONVERGENCE=',1PD12.5,' KEPEQ')
      EQ=ET
      ET=ET+DETN
      DO 1 I=1,100
      SINET=DSIN(ET)
      COSET=DCOS(ET)
      F=RN*(T+TSTAR)-ET+EPS*SINET
      DF=EPS*COSET-1.000
      ET=ET-F/DF
      DIF=DABS(F/DF)
      IF(DIF.LT.1.0D-10)GO TO 2
1  CONTINUE

```



```
WRITE(6,601)DIF
2 DETN=ET-ED
S(1)=A*(COSET-EPS)
S(2)=A*DSQRT(1.000-EPS*EPS)*SINET
S(3)=0.000
RETURN
END
//GO.SYSIN DD *
TEST RUN
TTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTTT
FFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFFF
10 6 4
34.0 39.0 0.01 0.1
.366447 .275514 .192315 -.141411 -.106223 -.072940
3318.76000 .0052056 .846765 6.0 1142.8
0.01 0.01 0.01
```